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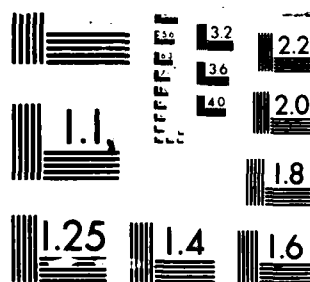
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**A FAMILY OF ALGORITHMS FOR THE
ESTIMATION OF THE PARAMETERS OF THE STABLE LAWS AND
THE PARAMETERS OF ATTRACTING STABLE LAWS**

by

**T. A. Delehanty
and
A. S. Paulson**



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DESCRIPTION AND PURPOSE

Potential application of the stable laws has long been hindered by the unavailability of generally available, well-documented algorithms. This paper removes this deficiency by presenting an algorithm for estimation of stable law parameters, with the goal of facilitating the application of stable laws in modeling and inference frameworks. The stable laws have steadily increased in importance to the statistical community since the paper of Mandelbrot (1963). Their role as the only laws possessing domains of attraction makes the stable laws an appealing probabilistic model, and they are capable of modeling a wide range of skewness, heavy tailedness, and central peakedness. Procedures for estimation of stable law parameters have been described by Mandelbrot (1963), DuMouchel (1971), Fama and Roll (1971), Paulson, Holcomb, and Leitch (1975), Koutrouvelis (1980,1981), Feuerverger and McDunnough (1981a,1981b), and Brockwell and Brown (1981). Because of the intractability of stable densities, attention has centered in recent years on Fourier-based procedures, using the empirical characteristic function. Such procedures should have an adaptive nature (Paulson, Holcomb, and Leitch, 1975; Paulson, Delehanty, and Brothers, 1982; Paulson and Delehanty, 1982).

~~This document~~
We present an iterative and adaptive algorithm for joint estimation of stable law parameters, using the empirical characteristic function. The algorithm is flexible in that either of two procedures may be selected, and subsets of the parameters may be allowed to vary freely, with others constrained or held constant. The statistical rationale for the procedures is described in the companion paper by Paulson and Delehanty (1982). The algorithm may also be used to provide informal estimates of the parameters

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estimation

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of the stable law to which a sample distribution is attracted.

THEORY AND NOTATION

Nondegenerate stable random variables X may be defined by the characteristic function

$$\phi(u) = E(\exp iuX) = \exp\{iu\mu - |\sigma u|^\alpha (1 + i\beta \operatorname{sgn}(u) \chi(u, \alpha))\}, \quad (1)$$

where $i^2 = -1$, $0 < \alpha \leq 2$, $|\beta| \leq 1$, $\sigma > 0$, and

$$\chi(u, \alpha) = \begin{cases} \tan \frac{\pi\alpha}{2}, & \alpha \neq 1 \\ \frac{2}{\pi} \log|u|, & \alpha = 1. \end{cases} \quad (2)$$

Here α , the characteristic exponent, is a measure of heavy tailedness and central peakedness, β is a skewness measure, σ is a scale parameter, and μ is a location parameter unless ($\alpha=1$, $\beta \neq 0$), when the function of location parameter is assumed by $\mu + \frac{2}{\pi} \beta \sigma \log \sigma$. The only stable laws whose densities are expressible in closed form are the Gaussian ($\alpha=2$, $\beta=0$), the Cauchy ($\alpha=1$, $\beta=0$), and the reciprocal of a χ^2 variate on one degree of freedom ($\alpha=\frac{1}{2}$, $\beta=-1$).

Let X_1, \dots, X_n be a stable random sample. The empirical characteristic function is

$$\phi_n(u) = n^{-1} \sum_{j=1}^n \exp(iuX_j). \quad (3)$$

Let $\psi(u) = \operatorname{Re} \phi(u) + i \operatorname{Im} \phi(u)$, $\psi_n(u) = \operatorname{Re} \phi_n(u) + i \operatorname{Im} \phi_n(u)$. Estimators interior to the parameter space can be viewed as zeros of the systems

Formulation A

$$\sum_{j=1}^q \frac{\partial \psi(u_j)}{\partial \theta} (\psi(u_j) - \psi_n(u_j)) w_j = 0, \quad (4)$$

or

Formulation B

$$\sum_{j=1}^q \sum_{k=1}^q \frac{\partial \psi(u_j)}{\partial \theta} K^{jk} (\psi(u_k) - \psi_n(u_k)) w_j w_k = 0, \quad (5)$$

for $\theta = \alpha, \beta, \sigma, \mu$. The grid $\{u_j | j=1, \dots, q\}$ is symmetric about zero but does not include the origin, and K^{jk} denotes the j, k element of the inverse matrix $(K_{jk})^{-1}$, where

$$\begin{aligned} K_{jk} &= n \operatorname{cov}(\psi_n(u_j), \psi_n(u_k)) \\ &= \operatorname{Re} \phi(u_j - u_k) + \operatorname{Im} \phi(u_j + u_k) - \psi(u_j) \psi(u_k). \end{aligned} \quad (6)$$

The weights $\{w_j | j=1, \dots, q\}$ also depend on the parameters $\alpha, \beta, \sigma, \mu$, and are described in the Numerical Method section. Both Formulations A and B represent modified, weighted χ^2 minimum procedures, corresponding to the respective objective functions

$$A: \quad S_n = \sum_{j=1}^q (\psi(u_j) - \psi_n(u_j))^2 w_j, \quad (7)$$

$$B: \quad Q_n = \sum_{j=1}^q \sum_{k=1}^q (\psi(u_j) - \psi_n(u_j)) w_j K^{jk} w_k (\psi(u_k) - \psi_n(u_k)). \quad (8)$$

The following points are critical for practical application:

- 1) The shapes of ψ and ψ_n are highly dependent on location and scale parameters, and so should be standardized;
- 2) The estimators are improved by making the gridpoints and weights depend on α and β ;

- 3) Since the procedures are adaptive ($\{u_j\}, \{w_j\}$ and $\{K_{jk}\}$ depend on unknown parameters), algorithms must be iterative;
- 4) Since α , β and σ are always constrained, each iteration involves solution of a nonlinear optimization problem with variable bound constraints.

Our procedures may therefore be summarized as follows, where a tilde indicates estimators, their values, or adaptively standardized quantities.

Begin with initial guesses for the parameters. At each iteration, compute and save $\{u_j\}$, $\{w_j\}$, possibly $\{K^{jk}\}$, and standardized empirical characteristic function values $\{\tilde{\psi}_n(u_j)\}$, based on the latest $(\tilde{\alpha}, \tilde{\beta})$. The objective S_n or Q_n is then minimized (an "optimization subproblem"), and cumulative location and scale estimates $(\tilde{\ell}, \tilde{s})$ are updated. Iteration stops when values of σ and μ minimizing S_n or Q_n are acceptably close to unity and zero, respectively.

Estimators whose values are not on a bound are asymptotically multivariate Gaussian distributed. The asymptotic covariance matrices $\underline{\Sigma}$ of the estimators are derived in Paulson and Delehanty (1982). The basic formula is

$$\underline{\Sigma}_i = \underline{H}_i^{-1} \underline{V}_i \underline{H}_i^{-1}, \quad i = A, B. \quad (9)$$

There are two particularly appealing ways to approximate $\underline{\Sigma}$. In "approximation (i)", expectations are approximated from the data: \underline{H} is computed by differencing the objective at the final optimum, and

$$\underline{V}_A = 4 \underline{D}^T \tilde{K} \underline{D}, \quad (10)$$

$$\underline{V}_B = 4 \underline{D}^T \underline{K}^{-1} \tilde{K}^{(w)} \underline{K}^{-1} \underline{D}. \quad (11)$$

Here

$$D_{j\theta} = \frac{\partial \tilde{\psi}(u_j)}{\partial \theta} w_j, \quad (12)$$

θ ranging over the parameters free of bounds,

$$\tilde{K}_{jk} = n^{-1} \sum_{m=1}^n (\tilde{\psi}(u_j) - \cos u_j \tilde{x}_m - \sin u_j \tilde{x}_m) (\tilde{\psi}(u_k) - \cos u_k \tilde{x}_m - \sin u_k \tilde{x}_m), \quad (13)$$

and

$$\tilde{K}_{jk}^{(w)} = \tilde{K}_{jk} w_j w_k. \quad (14)$$

By location and scale invariance, $(\tilde{\sigma}, \tilde{\mu})$ are set to $(1, 0)$ during these computations, and $\tilde{\psi}$ scaled. In "approximation (ii)", expectations are calculated analytically, so \tilde{K} replaces \tilde{K} in (10), (11), and (14), and factors of 2 are omitted. The expected Hessian has elements

$$H_{A\theta\theta'} = \sum_{j=1}^q \frac{\partial \tilde{\psi}(u_j)}{\partial \theta} \frac{\partial \tilde{\psi}(u_j)}{\partial \theta'} w_j, \quad (15)$$

$$H_{B\theta\theta'} = \underline{D}^T \underline{K}^{-1} \underline{D},$$

where θ and θ' range over free parameters.

To analyze domains of attraction, we use what we refer to as the k -sum procedure. If k is a positive integer, the power

$$\phi_n^k(u) = n^{-k} \sum_{j_1=1}^n \cdots \sum_{j_k=1}^n \exp(iu(x_{j_1} + \cdots + x_{j_k})) \quad (16)$$

is the characteristic function corresponding to the k^{th} convolution

power of the empirical distribution function $F_n(x)$, and can be interpreted as empirical characteristic function of all possible k -sums

$\{x_{j_1} + \cdots + x_{j_k}\}$, sampling with replacement from $F_n(x)$. We add real and

imaginary parts and standardize, giving $\tilde{\psi}_n^k(u)$, say, and estimate $(\tilde{\alpha}_k,$

$\tilde{\beta}_k, \tilde{\sigma}_k, \tilde{\mu}_k)$ for different values of k . If the sample distribution is

attracted to a stable law with parameters $(\alpha, \beta, \sigma, \mu)$, the sequence of normalized estimators $(\tilde{\alpha}_k, \tilde{\beta}_k, \tilde{\sigma}_k/k^{1/\tilde{\alpha}_k}, \tilde{\mu}_k/k)$, for reasonable values of k , should approach $(\alpha, \beta, \sigma, \mu)$. In particular, a rapid rise in $\{\tilde{\alpha}_k\}$ may indicate that a stable model is not appropriate, a possible alternative being a mixture of finite variance components with differing scale parameters.

The k -sum procedure can thus be used in a sensitivity analysis, to examine how well the data support the stability assumption. Other possible tools for sensitivity analysis are varying the mechanism (to be described below) underlying the weights $\{w_j\}$, and comparing approximations (i) and (ii) of the estimated asymptotic covariance matrix, provided n is large enough for approximation (i) to be accurate.

NUMERICAL METHOD

The main computational task required is solution of bound-constrained nonlinear optimization problems. Although Formulations A and B lead to nonlinear least squares problems, current algorithms for nonlinear least squares do not allow constraints (Hiebert, 1981). Numerical Algorithms Group (NAG) subroutine E04KBF (NAG, 1981) is used for optimization. E04KBF is a quasi-Newton procedure, requiring an objective function and analytical first partial derivatives. It is substantially faster than the gradient projection routine used by Paulson, Holcomb and Leitch (1975), although the latter is very reliable. The other complicated numerical procedure required is inversion of a positive definite symmetric matrix $(K, H \text{ or } \tilde{H})$, for which NAG subroutine F01ABF is used. Various NAG utility procedures are also used, see Auxiliary Algorithms. The use of NAG

procedures inhibits transportability in that the algorithm, as presented, is only usable at installations having the NAG Library. However, listings of rapid, high-quality algorithms for constrained optimization have not appeared in the literature (see Chambers, 1977, pp. 159-160; the situation described there has not improved). Given that E04KBF is used, reliance on additional NAG Library procedures is expedient.

We require a minimum of $q=20$ gridpoints $\{u_j\}$, and prefer $q=20$ or 40 , since they are reasonable values in practice, and have been tested extensively. Only the positive gridpoints are explicitly required, due to symmetry of the grid and the Hermitian property of characteristic functions. They are computed as follows: An endpoint U is chosen as $3, \tilde{\alpha} \geq 1.8$; $3.3, 1.8 > \tilde{\alpha} \geq 1.7$; $3.6, 1.7 > \tilde{\alpha} > 1$; $5, \tilde{\alpha}=1$; $4, 1 > \tilde{\alpha} \geq .9$; $5, .9 > \tilde{\alpha} \geq .8$; $7, .8 > \tilde{\alpha} \geq .6$; $10, \tilde{\alpha} < .6$. An inner number I of points is selected close to the origin: $I=2$ if $q < 30$ and 3 if $q \geq 30$, 1 being subtracted if $\tilde{\alpha} \leq .5$. The inner I points are spaced as follows: if $\tilde{\alpha} > 1$, $\frac{1}{2}$ the " α -optimal" values of Feuerverger and McDunnough (1981b) for the nearest (larger) α are used; if $\tilde{\alpha} \leq 1$, the first I points giving $q/2$ equal increments of $\log(u + \alpha^{*3})$ between 0 and U are multiplied by $\frac{1}{2} \alpha^{*2}$ ($\alpha^* = \max(\tilde{\alpha}, .3)$). The remaining points are logarithmically spaced out to U : if $\tilde{\alpha} > 1$, the function $\log(1 + u/2)$ is used, and if $\tilde{\alpha} \leq 1$, $\log(u + \alpha^{*3})$ is used. This rather complicated ad hoc scheme was developed through graphical inspection of $\tilde{\psi}(u)$ and $\tilde{\psi}_n(u)$, comparisons of asymptotic efficiencies, and parameter estimation for real and simulated data. No claims of optimality are made, but the scheme provides high efficiencies if efficiency is preferred, or good matches between $\tilde{\psi}$ and $\tilde{\psi}_n$ if curve fitting is preferred. The point of stratified and logarithmic spacing is

to emphasize u values near the origin. Details when $\tilde{\alpha} \leq 1$ reflect the fact that $\psi(u)$ has a sharp cusp near the origin, but decays slowly. The stepwise nature of the scheme is not deemed a serious drawback.

The weights $\{w_j\}$ are computed as follows:

Under Formulation A,

$$w_j = \frac{|\phi(u_j)|^{2\lambda}}{K_\tau(u_j, u_j)} = \frac{\exp(-2\lambda|u_j|^\alpha)}{K_\tau(u_j, u_j)}, \quad (17)$$

and under Formulation B,

$$w_j = |\phi(u_j)|^\lambda = \exp(-\lambda|u_j|^\alpha), \quad (18)$$

where λ and τ are supplied by the user, $0 \leq \tau \leq 1$,

$$K_\tau(u, u) = 1 + \tau(\text{Im } \tilde{\phi}(2u) - \tilde{\psi}^2(u)), \quad (19)$$

and λ is recommended nonnegative. Rationale for these weights, and some corresponding asymptotic efficiencies, are in Paulson and Delehanty (1982). We recommend $\tau=1$ under Formulation A. Under Formulation B, it is convenient to let $\tau \geq 0$ represent a fraction of the average diagonal element by which to inflate K , giving a matrix we shall call A . We have only found this inflation necessary if $\tilde{\alpha}$ is very close to two, when $\tau=0.01$ suffices.

To use the quantity λ as a tool for sensitivity analysis, we interpret it as a damping factor, lessening the effects of noise in $\tilde{\psi}_n(u)$ for larger $|u|$. If the data are truly stable and the sample size is fairly large (say 150 or more), estimates should change little as λ varies, say, from 0 to 1. Large discrepancies in the estimates for different values of λ indicate problems with the data or the stable assumption or both. It may not be easy to isolate the difficulty but further study is

definitely required.

For the k-sum procedure, $k > 1$, the situation regarding gridpoints and weights changes. Tests so far indicate that when $\tilde{\alpha} > 1$, Formulation A, with gridpoints equispaced from 0 to U, gives better results than "efficient configurations" used for $k=1$. Reasons for this are unclear. A possible explanation is that when $\tilde{\alpha} > 1$ and $k > 1$, $\tilde{\psi}_n^k(u)$ is so smooth that estimation is practically equivalent to deterministic curve fitting, and implicit or explicit emphasis on gridpoints near the origin neglects important curvature for large $|u|$. Accordingly, when $k > 1$ and $\tilde{\alpha} > 1$, we equispace gridpoints and set all weights to 1. When $\tilde{\alpha} \leq 1$, $\tilde{\psi}_n^k(u)$ has a sharp cusp near the origin and remains a jagged curve as k increases, due to the presence of very large observations. In this situation, we set all weights to 1 and use basically the same gridpoint scheme as when $k=1$, omitting only multiplication of the inner points by $\alpha^{*2/4}$. In either case, Formulation A is recommended.

An important question is how large k may be taken. Equation (16) suggests that we cannot expect to take k arbitrarily large. There seems to be a tendency for $\tilde{\alpha}$ to increase and $\tilde{\beta}$ to drift if k is too large, though this may be partially due to suboptimal gridpoints or weighting. It appears that when n is large, say 500 or more, k may safely be taken up to 20. Care is required for smaller n , and when α is small or very near two.

Implicit standardization is carried out as follows. Let k be a positive integer, and $(\tilde{\ell}, \tilde{s})$ cumulative location and scale estimates. Then

$$\tilde{\psi}_n^k(u_j) = \rho_{jk}(\cos \gamma_{jk} + \sin \gamma_{jk}), \quad (20)$$

where

$$\rho_{jk} = |\phi_n(u_j/\tilde{s})|^k \quad (21)$$

and

$$\gamma_{jk} = k \arg \phi_n(u_j/\tilde{s}) - \tilde{\ell} u_j/\tilde{s}. \quad (22)$$

No problems of principal values arise, and complex arithmetic is not used. The FORTRAN mathematical library function ATAN2 computes arguments.

The estimator $\tilde{\alpha}$ may be bounded in (closed) subintervals of $[\delta, 1-\epsilon]$, $[1, 1]$, or $[1+\epsilon, 2]$ unless $\tilde{\beta}$ is fixed at 0, when $[\delta, 2]$ is possible (δ and ϵ are small positive numbers), while $\tilde{\beta}$ may be bounded in subintervals of $[-1, 1]$. Estimators $\tilde{\sigma}$ and $\tilde{\mu}$ may be constrained arbitrarily in $[\delta, \infty)$ and $(-\infty, \infty)$, respectively, unless $\tilde{\alpha}$ is fixed at 1 and $\tilde{\beta}$ is not fixed at 0, when $\tilde{\mu}$ cannot be constrained. Bounds on σ and μ are internally set for use in subproblems. These bounds must be wide enough to allow the "true values" to be found, but narrow enough to deter straying into undesirable regions, particularly $\sigma \rightarrow \infty$, $|\mu| \rightarrow \infty$. The ad hoc bounds of $[-5, 5]$ for μ and $[0.2, 5]$ for σ work well in practice. If $\tilde{\sigma}$ or $\tilde{\mu}$ are initially constrained, their internal bounds are adaptively modified, see the Algorithm for description.

Initial guesses for the parameters are required. We do not find their specification particularly important, provided $\tilde{\alpha}$ is on the correct side of 1 in the nonsymmetric case. We have used the median and semi-interquartile range as guesses for $\tilde{\mu}$ and $\tilde{\sigma}$, and averages of upper and lower bounds for $\tilde{\alpha}$ and $\tilde{\beta}$. If $\tilde{\alpha}$ is anticipated less than 1.2, say, it is

worthwhile to put more effort into initial guesses, since fewer iterations will be required (the semi-interquartile range will overestimate σ , and if α is near but different from 1, the median is nearer $\mu - \frac{2}{\pi} \beta \sigma \log \sigma$ than μ).

Convergence is judged by a tolerance on subproblem solutions, $\max(|\sigma^{(m)} - 1|, |\mu^{(m)}|)$, (or $\max(|\alpha^{(m)} - \alpha^{(m-1)}|, |\beta^{(m)} - \beta^{(m-1)}|)$ if $\bar{\sigma}$ and $\bar{\mu}$ are fixed), with a maximum allowable number of iterations. Attainable tolerances depend on n , but more strongly on the underlying parameters. If $\bar{\alpha}$ is near two, $\bar{\psi}_n$ is very smooth and stringent tolerances such as 10^{-6} may be attained. If $\bar{\alpha} \leq 1.2$, $\bar{\psi}_n$ has many small oscillations due to large observations, and, especially for smaller samples, it may be preferable to terminate after a fixed number of iterations. Good estimates are generally obtained within five iterations, fewer if initial guesses are good; if stringent tolerances are required, or for difficult problems (skewed distributions with $0.9 \leq \alpha \leq 1.1$) more may be required. Convergence is typically slower under Formulation B, since the weighting mechanism is more complicated.

Approximation of asymptotic covariance matrices requires little description. We note that for approximation (i) and the q values we use, it is faster to define a vector

$$\delta_j^T = (\bar{\psi}(u_1) - \cos u_1 \bar{x}_j - \sin u_1 \bar{x}_j, \dots, \bar{\psi}(u_q) - \cos u_q \bar{x}_j - \sin u_q \bar{x}_j), \quad (23)$$

and cumulate

$$\bar{V} = 4n^{-1} \sum_{j=1}^n (\bar{D}^T \delta_j) (\bar{D}^T \delta_j)^T \quad (24)$$

under Formulation A, or

$$\tilde{V} = 4n^{-1} \sum_{j=1}^n (\underline{D}^T \underline{A}^{-1} \delta_j)(\underline{D}^T \underline{A}^{-1} \delta_j)^T \quad (25)$$

under Formulation B, than to cumulate \tilde{K} . The matrix \underline{D} is computed by the function/gradient subroutine. E04KBF returns an approximate Hessian, which could conceivably be used for \tilde{H} in approximation (i). Rather often, however, E04KBF will terminate with its failure indicator set to 3 and the Hessian set to the identity matrix, even though the optimum may be reliable. It is therefore simpler to compute \tilde{H} by differencing. The following procedure is used: Set an initial Hessian to 0, and the steplength to 10^{-3} . Successively divide the steplength by $\sqrt{10}$ and approximate the Hessian by differencing; three-point differencing for the diagonal, and four-point for off-diagonal elements. Compare elements of successive approximations by maximum relative or absolute differences, according as the element of the latest approximation exceeds 1 in absolute value or not. A tolerance of 10^{-6} is used for this convergence criterion. If convergence has not occurred with a steplength of 10^{-5} , the result with steplength 10^{-4} is used.

Approximation (i) of the asymptotic covariance matrix is rather expensive to compute. It should not be computed for smaller sample sizes, as it implicitly involves estimation of $\frac{1}{2}q(\frac{1}{2}q+1)$ covariances.

Following is an informal description, in Algorithm form, of the basic routine STABLE. Approximate asymptotic covariance matrices may also be computed, but this presents no logical difficulties, so is omitted.

Algorithm

Produces estimates $(\tilde{\alpha}, \tilde{\beta}, \tilde{\sigma}, \tilde{\mu})$ for k -sums ($k \geq 1$), based on a sample (x_1, \dots, x_n) .

Input parameters: $k, n, \{x_j\}, q, \lambda, \tau$, Formulation (A or B), convergence tolerance ϵ , maximum number M of iterations, and flags whether $\tilde{\sigma}$ and $\tilde{\mu}$ are constrained.

Input/output parameters: $(\tilde{\alpha}, \tilde{\beta}, \tilde{\sigma}, \tilde{\mu})$ are initial guesses on entry and estimates on exit, $(\alpha_L, \beta_L, \sigma_L, \mu_L)$ and $(\alpha_u, \beta_u, \sigma_u, \mu_u)$ are lower bounds. The (σ, μ) bounds are changed, but restored on exit. In the special case where $\tilde{\alpha}$ is fixed at 1, $\tilde{\mu}$, on entry, is the initial guess for location $\mu + \frac{2}{\pi} \beta \log \sigma$.

Auxiliary quantities $\tilde{\ell}$ and \tilde{s} are cumulative location and scale estimates. Entry values of $(\sigma_L, \sigma_u, \mu_L, \mu_u)$ are stored in (b_1, b_2, b_3, b_4) . On entry and exit, $(\tilde{\sigma}, \tilde{\mu}, \sigma_L, \mu_L, \sigma_u, \mu_u)$ are normalized.

S1 [Initialize.]

Set $\tilde{\ell} \leftarrow k\tilde{\mu}$, $\tilde{s} \leftarrow k^{1/\tilde{\alpha}} \tilde{\sigma}$, $m \leftarrow 0$.

Save $(b_1, b_2, b_3, b_4) \leftarrow (\sigma_L, \sigma_u, \mu_L, \mu_u)$.

if $\tilde{\mu}$ is unconstrained then set $\mu_L \leftarrow -5$, $\mu_u \leftarrow 5$;

else if $\tilde{\mu}$ is fixed then set $\mu_L \leftarrow \mu_u \leftarrow 0$;

else set $\mu_L \leftarrow k\mu_L$, $\mu_u \leftarrow k\mu_u$.

if $\tilde{\sigma}$ is unconstrained then set $\sigma_L \leftarrow 0.2$, $\sigma_u \leftarrow 5$;

else if $\tilde{\sigma}$ is fixed then set $\sigma_L \leftarrow \sigma_u \leftarrow 1$;

else set $\sigma_L \leftarrow k^{1/\tilde{\alpha}} \sigma_L$, $\sigma_u \leftarrow k^{1/\tilde{\alpha}} \sigma_u$.

S2 [Looping point for iteration; save adaptive quantities for subproblem.]

Increment $m \leftarrow m + 1$.

Save $\tilde{\alpha}^{(m-1)} \leftarrow \tilde{\alpha}$, $\tilde{\beta}^{(m-1)} \leftarrow \tilde{\beta}$.

if $\tilde{\sigma}$ is constrained but not fixed then set $\sigma_L \leftarrow \max(0.2, b_1/\tilde{s})$,
 $\sigma_U \leftarrow \min(5, b_2/\tilde{s})$.

if $\tilde{\mu}$ is constrained but not fixed then set $\mu_L \leftarrow \max(-5, (b_3 - \tilde{\ell})/\tilde{s})$,
 $\mu_U \leftarrow \min(5, (b_4 - \tilde{\ell})/\tilde{s})$.

Set $\tilde{\sigma} \leftarrow \tilde{\sigma} + 1$, $\tilde{\mu} \leftarrow 0$.

Compute and save positive gridpoints $\{u_j \mid j = q/2+1, \dots, q\}$,
 weights $\{w_j \mid j=1, \dots, q\}$, and standardized empirical characteristic
 function values $\{\tilde{\psi}_n^k(u_j) \mid j=1, \dots, q\}$.

if Formulation B then compute and invert A.

S3 [Subproblem.]

Solve the optimization problem

$$\min \sum_{j=1}^q (\psi(u_j) - \tilde{\psi}_n^k(u_j))^2 w_j \quad (\text{Formulation A})$$

or

$$\min \sum_{i=1}^q \sum_{j=1}^q w_i (\psi(u_i) - \tilde{\psi}_n^k(u_i)) A^{ij} (\psi(u_j) - \tilde{\psi}_n^k(u_j)) w_j \quad (\text{Formulation B}),$$

yielding new $(\tilde{\alpha}, \tilde{\beta}, \tilde{\sigma}, \tilde{\mu})$.

S4 [Update and test convergence.]

Set $\tilde{\ell} \leftarrow \tilde{\ell} + \tilde{s} \tilde{\sigma}$.

if $\tilde{\alpha}=1$ then set $\tilde{\ell} \leftarrow \tilde{\ell} + \frac{2}{\pi} \tilde{s} \tilde{\beta} \tilde{\sigma} \log \tilde{\sigma}$.

Set $\tilde{s} \leftarrow \tilde{s} \tilde{\sigma}$.

if $\tilde{\sigma}$ and $\tilde{\mu}$ are fixed then error $\leftarrow \max (|\tilde{\alpha} - \tilde{\alpha}^{(m-1)}|, |\tilde{\beta} - \tilde{\beta}^{(m-1)}|)$;

else error $\leftarrow \max (|\tilde{\sigma} - 1|, |\tilde{\mu}|)$.

if error $\geq \epsilon$ and $m < M$ then go to S2.

S5 [Final estimates.]

Set $(\sigma_L, \sigma_U, \mu_L, \mu_U) \leftarrow (b_1, b_2, b_3, b_4)$.

if $\tilde{\alpha} = 1$ then set $\tilde{\ell} \leftarrow \tilde{\ell} - \frac{2}{\pi} \tilde{\beta} \tilde{s} \log \tilde{s}$.

Set $\tilde{\mu} \leftarrow \tilde{\mu} / k$, $\tilde{\sigma} \leftarrow \tilde{s} / k^{1/\tilde{\alpha}}$.

STRUCTURE

SUBROUTINE STABLE (X,N,MODE,KSUM,XLAM,TAU,NPTS,TOL,MAXIT,XL,XB,XH,NPAR,
ISCLBD,LOCBND,ICOV,VCV1,VCV2,WORK,LWORK,IWORK,LIWORK,IFALT)

Formal parameters

X	Real array (N)	input:	sample
N	Integer	input:	sample size
MODE	Integer	input:	formulation; if zero, then Formulation B is used, else Formulation A
KSUM	Integer	input:	convolution power k
XLAM	Real	input:	λ
TAU	Real	input:	τ
NPTS	Integer	input:	q
TOL	Real	input:	convergence tolerance
MAXIT	Integer	input:	maximum allowable number of iterations
XL	Real array (NPAR)	input:	lower bounds for parameters ($\alpha, \beta, \sigma, \mu$); the third and fourth elements change during execution
		output:	input values are restored
XH	Real array (NPAR)	input:	upper bounds for parameters; the third and fourth elements change during execution
		output:	input values are restored
NPAR	Integer	input:	number of parameters (4)
ISCLBD	Integer	input:	flag if $\bar{\sigma}$ is constrained: if negative, $\bar{\sigma}$ is fixed at XB(3); if zero, $\bar{\sigma}$ is free to vary and initial values of XL(3) and XH(3) are irrele- vant; if positive, $\bar{\sigma}$ is constrained in [XL(3),XH(3)]

LOCBND	Integer	input:	flag if $\bar{\mu}$ is constrained: if negative, $\bar{\mu}$ is fixed at XB(4); if zero, $\bar{\mu}$ is free to vary and initial values of XL(4) and XH(4) are irrelevant; if positive, $\bar{\mu}$ is constrained in [XL(4),XH(4)]
ICOV	Integer	input:	flag for computation of covariance matrices: if negative, neither approximation (i) nor (ii) is computed; if zero, both approximations are computed; if positive, only approximation (ii) is computed
VCV1	Real array(NPAR,NPAR)	output:	covariance matrix approximation (i) if requested; the strict lower triangle contains correlations, the upper triangle contains covariances (times n); if a parameter is on a bound, the corresponding elements are zero
VCV2	Real array(NPAR,NPAR)	output:	covariance matrix approximation (ii) if requested; the strict lower triangle contains correlations, the upper triangle covariances (times n); if a parameter is on a bound, the corresponding elements are zero
WORK	Real array (LWORK)	workspace:	
		output:	some elements may be of interest on output (see Restrictions)
LWORK	Integer	input:	
IWORK	Integer array(LIWORK)	workspace:	
		output:	some elements may be of interest on output (see Restrictions)
LIWORK	Integer	input:	
IVNIT	Integer	input:	if positive, unit number for output (see Additional Comments); if zero or negative, no output is produced
IFAUULT	Integer	output:	failure indicator

Failure indicators

IFAUULT = 0 indicates success. Nonzero values of IFAULT are due to two types of errors. The first type is input errors, detected in STABLE; IFAULT will be

- 1 if MAXIT \leq 0;
- 2 if N \leq 50 (see Restrictions);
- 3 if KSUM \leq 0;
- 4 if $\tau < 0$ or $\tau > 1$ and MODE \neq 0;
- 5 if NPTS \leq 20 or mod(NPTS,2) \neq 0;
- 6 if TOL \leq 0;
- 7 if NPAR \neq 4;
- 8 if insufficient workspace was allotted (see Restrictions);
- 9 if improper bounds were supplied. The following conditions cause this failure:
 $XL(i) > XB(i)$ or $XL(i) > XH(i)$ or $XB(i) > XH(i)$, $i=1,2$
 $XL(1) \leq 0$ or $XH(1) > 2$
 $XL(1) < -1$ or $XH(1) > 1$
 $(XL(2) \neq 0$ or $XH(2) \neq 0)$ and $(XL(1) < 1$ and $XH(1) \geq 1$
or $XL(1) \leq 1$ and $XH(1) > 1)$
 $XB(3) \leq 0$
ISCLBD \neq 0 and $(XL(3) > XB(3)$ or $XL(3) > XH(3)$ or $XB(3) > XH(3))$
ISCLBD $<$ 0 and $XL(3) \neq XH(3)$
LOCBND \neq 0 and $(XL(4) > XB(4)$ or $XL(4) > XH(4)$ or $XB(4) > XH(4))$
LOCBND $<$ 0 and $XL(4) \neq XH(4)$
LOCBND \neq 0 and $XL(1)=XH(1)=1$ and $(XL(2) \neq 0$ or $XH(2) \neq 0)$.

On input errors, STABLE terminates immediately, without performing any computations. The second type of error occurs after some computation.

IFAUULT will be

- 10 if A was found numerically non positive definite;
- 11 if A was found ill-conditioned;
- 12 if too many function evaluations were required during solution of a subproblem;
- 13 if iteration converged, but the most recent E04KBF fault indicator was 3 and internal checks were not met. These checks are
(i) $\|G\|^2 < 10 * X02AAF(DUMMY)$, and
(ii) $K < 1/\|G\|$, as recommended by E04KBF documentation, where $\|G\|$ is the norm of the projected gradient and K the estimated condition number of the projected Hessian matrix;

- 14 if there were repeated problems with overflow in the Cholesky factors of the projected Hessian;
- 15 if iteration converged, but the most recent E04KBF fault indicator was 5 and internal checks were not met;
- 16 if convergence did not occur in MAXIT iterations;
- 17 if convergence did not occur in MAXIT iterations, the most recent E04KBF fault indicator was 3, and internal checks were not met;
- 18 if convergence did not occur in MAXIT iterations, the most recent E04KBF fault indicator was 5, and internal checks were not met.

Conditions IFAULT=10 and 11 are detected in SETECF (they are caused by τ being too small under Formulation B), the remainder in STABLE.

IWORK(2) and IWORK(3) (see Restrictions) are failure indicators for asymptotic covariance matrix versions (i) and (ii) respectively. Zero indicates success, 1 that H was non positive definite, and 2 that H was ill-conditioned, the failures detected in SETVCV. If IFAULT=1-12 or 14, covariance matrices are not computed, and their fault indicators are set to the corresponding value of IFAULT.

Auxiliary algorithms

The user has only to call STABLE. Auxiliary procedures fall into two groups: those supplied here, and NAG Library procedures. The following subroutines are supplied:

SUBROUTINE GRIDWT(PAR,NPAR,XLAM,TAU,PTS,NPTS2,WT,NPTS,MODE,KSUM): computes gridpoints and weights;

SUBROUTINE CHARFN(U,PAR,NPAR,RE,XIM): computes real and imaginary parts of standard stable characteristic function $\phi(u)$;

SUBROUTINE FUNCT(IFLAG,N,XC,FC,GC,IW,LIW,W,LW): objective function/gradient evaluation;

SUBROUTINE SETECF(X,N,PAR,NPAR,MODE,TAU,SIGMA,XMU,KSUM,IA,NPTS2,NPTS,PTS,ECF,A,AINV,WORK,IFAILT): computes standardized empirical characteristic function values $\tilde{\psi}_n^k(u)$, computes and inverts A under Formulation B;

SUBROUTINE MONIT(N,XC,FC,GC,ISTATE,GPJNRM,COND,POSDEF,NITER,NF,IW,LIW,W,LW): monitors the progress of E04KBF;

SUBROUTINE VARIAB(ICOV,X,N,PAR,NPAR,MODE,SIGMA,XMU,ISUB,NVAR,PTS,NPTS2,WT,ECF,NPTS,DERIV,WORK,HOLD,A,IA,AINV,VCV1,VCV2,H,NVAR1,V,IW,LIW,W,LW,IFAIL1,IFAIL2): computes approximate asymptotic covariance matrices;

SUBROUTINE VMATRX (X,N,MODE,XMU,SIGMA,PTS,NPTS2,WT,ECF,WORK,NPTS,DERIV,V,HOLD,NVAR): computes \hat{Y} for version (i) of asymptotic covariance matrix;

SUBROUTINE DAPROD(FAC1,IFAC1,NPTS,FAC2,WORK,NVAR): auxiliary matrix multiplication for VARIAB;

SUBROUTINE HVPROD(FAC1,IFAC1,NVAR,FAC2,NPTS,VH,IVH): auxiliary matrix multiplication for VARIAB;

SUBROUTINE SETVCV(ISUB,NVAR,H,NVAR1,V,WORK,VCV,NPAR,SIGMA,IFAILT): auxiliary routine for VARIAB;

SUBROUTINE HESDIF(PAR,NPAR,ISUB,H,SAVE1,SAVE2,NVAR,IW,LIW,W,LW): computes an approximate Hessian by differencing for version (i) of asymptotic covariance matrix.

The following NAG Library procedures are used:

REAL FUNCTION X02AAF(DUMMY): returns the smallest positive ϵ such that $1.0 + \epsilon > 1.0$;

SUBROUTINE E04KBF(N,FUNCT,MONIT,IPRINT,LOCSCH,INTYPE,MINLIN,MAXCAL,ETA,XTOL,STEPMX,FEST,IBOUND,BL,BU,X,HESL,LH,HESD,ISTATE,F,G,IW,LIW,W,LW,IFAIL): solves optimization problems. Control parameters are set as follows:

```

IPRINT  =  0
LOCSCH  =  .TRUE.
INTYPE  =  3 for subproblems after the first if parameters which are
           not fixed are not on bounds, else 0
MINLIN  =  NAG Library routine E04LBS
MAXCAL  =  400
ETA     =  0.9
XTOL    =  10.0* $\sqrt{\text{X02AAF(DUMMY)}}$  explicitly, so it is available on exit
STEPMX  =  0.25
FEST    =  0.0
IBOUND  =  0 ;

```

SUBROUTINE F01ABF(A,IA,N,B,IB,Z,IFAIL): inverts the positive definite symmetric matrix A;

SUBROUTINE F01CAF(A,M,N,IFAIL): sets matrix A to zero;

SUBROUTINE F01CMF(A,LA,B,LB,M,N): copies elements of matrix A into matrix B;

SUBROUTINE F01CKF(A,B,C,N,IP,M,Z,IZ,IOPT,IFAIL): matrix multiplication $A=BC$, where B or C may be overwritten.

RESTRICTIONS

We require the sample size N at least 50, since for smaller samples $\tilde{\psi}_n(u)$ is not generally sufficiently smooth to allow accurate estimation. Since $\tilde{\alpha}$ and $\tilde{\beta}$ are bounded in the narrow ranges $(0,2]$ and $[-1,1]$ and have standard errors decreasing as $N^{-1/2}$, it is preferable to have $N \geq 100$. For N less than 150, say, relatively large values of λ may be preferred, to damp out noise in $\tilde{\psi}_n(u)$. We further require $NPTS \geq 20$.

Extended work vectors WORK and IWORK are required, in order to communicate information to FUNCT and MONIT without using COMMON blocks. To aid readers who may wish to adapt the algorithm to installations not having the NAG Library, we describe the use of these work vectors.

The required length of WORK is $10 + 11*NPAR + NPAR*(NPAR-1)/2 + (3+NPAR)*NPTS + NPTS + NPTS/2$ if $MODE \neq 0$, with an additional $NPTS*(2*NPTS+1)$ required if $MODE=0$. Some sample lengths are

<u>MODE</u>	<u>NPTS=20</u>	<u>NPTS=40</u>
0	1030	3600
nonzero	210	360

The subvector W is passed to E04KBF, FUNCT, and MONIT.

WORK starting pointW starting pointElementsUsed for

1	-	1	Convergence criterion
2	-	1	Objective function value on exit from E04KBF
3	-	NPAR	Projected gradient on exit from E04KBF
(other addresses internally computed)	-	1	Old α value for convergence testing
	-	1	Old β value for convergence testing
	-	1	α lower bound on entry
	-	1	α upper bound on entry
	-	1	μ lower bound on entry
	-	1	μ upper bound on entry
	-	NPAR*(NPAR-1)/2	HESL factor for E04KBF
	-	NPAR	HESD factor for E04KBF; workspace for VARIAB
	-	9*NPAR	E04KBF workspace; broken up into H and V matrices and used as workspace in VARIAB
	1 (other addresses internally computed)	1	On exit from E04KBF, estimated condition number of projected Hessian
		1	On exit from E04KBF, norm of projected gradient
		NPTS/2	Positive gridpoints PTS
		NPTS	Weights WT
		NPTS	Empirical characteristic function values ECF, workspace in VARIAB
		NPAR*NPTS	Partial derivatives of $\psi(u)$ at gridpoints, workspace in VARIAB
		NPTS	Workspace for FUNCT and VARIAB
		(NPTS+1)*NPTS	If MODE=0, used for A matrix (the extra row is required by NAG Library routine F01ABF); workspace in VARIAB
		NPTS*NPTS	If MODE=0, used for A^{-1} ; workspace in VARIAB

The required length of IWORK is $7 + \text{NPAR}$. The subvector IW is passed to E04KBF, FUNCT, and MONIT.

<u>IWORK starting point</u>	<u>IW starting point</u>	<u>Elements</u>	<u>Use for</u>
1	-	1	Iteration count
2	-	NPAR	ISTATE vector for E04KBF, workspace for VARIAB.
(other addresses internally computed)			If covariance matrices are requested, on exit IWORK(2) contains a fault indicator for approximation (i), IWORK(3) contains a fault indicator for approximation (ii), and IWORK(4) contains the number of iterations required to compute the approximate Hessian for approximation (i)
	1	2	Workspace for E04KBF, HESDIF
	3	1	Stores MODE
	4	1	Stores output unit number IUNIT
	5	1	Stores NPTS
	6	1	Stores 1 less than the address of PTS(1) in W

PRECISION

Double precision will be required on computers with 32 bit wordlength. The precision used by the local NAG Library implementation should be adequate. To change the precision:

- change all REAL declarations to DOUBLE PRECISION;
- replace constants by double precision versions, constants $\frac{\pi}{2}$, $\frac{2}{\pi}$, $\sqrt{10}$ typed in to machine accuracy;
- declare NAG Library function X02AAF as DOUBLE PRECISION;
- change the precision of FORTRAN library functions, i.e., ABS to DABS, ATAN2 to DATAN2, SIGN to DSIGN, etc. FLOAT(I) can be replaced by DBLE(FLOAT(I)).

If extremely large observations are present in the sample, there may be a loss of significant figures when computing sines and cosines in SETECF and VMATRX. This should not occur when real data is used, but can be a problem with simulated data for small α .

TIME

Execution times depend on the quality of initial guesses and properties of the real data used, and vary somewhat throughout the parameter space. As a rough guide, we give some statistics for simulated data, using a moderately difficult situation with $\alpha > 1$. Tables 1a and 1b provide approximate running times for Formulation A, $q=40$, and Formulation B, $q=20$, $n=100, 200, 500, 1000, 2500$. Timing starts upon entry to STABLE. Samples from $S(1.3, -.5, 3, 15)$ were generated using the method of Chambers, Mallows, and Stuck (1976). Initial guesses for $\alpha, \beta, \sigma, \mu$ in all cases were $1505 = \frac{1}{2}(1.01+2)$, 0, $\frac{1}{2}(x_{.75} - x_{.25})$, and $x_{.5}$, the sample median, respectively. Because of skewness, the median is not a good estimator of μ in this case. Five iterations were used. Time required to compute asymptotic covariance matrices includes approximations (i) and (ii), except where noted. Timings are for a double precision version of the algorithm, compiled by the IBM FORTRAN H Extended compiler, and run on an IBM 370/3033.

The following qualitative points are clear from this rather restricted set of timings. There is a substantial overhead, which may crudely be assumed fixed, associated with nonlinear optimization, although E04KBF solves the optimization subproblems rapidly. For large samples, run time is dominated by evaluation of the empirical characteristic function, and thus is asymptotically linear in n for a fixed number of iterations.

Table 1a

Timings for Formulation A, $q=40$, on Simulated Samples
from $s(1.3, -.5, 3, 15)$; $\lambda=1$ for $n \leq 200$ and $.5$ for $n > 200$, $\tau=1$

<u>n</u>	<u>Iterations</u>	<u>Estimation time (sec)</u>	<u>Convergence criterion</u>	<u>Covariance matrix time</u>
100	5	0.7	5.4(-4)	0.1*
200	5	1.0	2.3(-2)	0.1*
500	5	1.8	1.8(-4)	0.8
1000	5	3.2	3.3(-5)	1.2
2500	5	7.5	4.8(-6)	2.5

*Sample size too small to compute approximation (i), only
approximation (ii) computed.

Table 1b

Timings for Formulation B, $q=20$, on Simulated Samples
from $s(1.3, -.5, 3, 15)$; $\lambda=\tau=0$

<u>n</u>	<u>Iterations</u>	<u>Estimation time (sec)</u>	<u>Convergence criterion</u>	<u>Covariance matrix time</u>
100	5	0.9	1.2(-2)	0.3
200	5	1.2	3.2(-2)	0.4
500	5	1.6	1.8(-4)	0.5
1000	5	2.3	5.7(-5)	0.7
2500	5	4.4	1.5(-4)	1.4

Approximation (ii) of the asymptotic covariance matrix is quite easy to compute, while approximation (i) is highly time-consuming.

For fixed $k > 1$ with the k -sum procedure, one iteration generally suffices, provided estimates from the nearest value of k are used, and the estimates don't change much. For mixtures of very different distributions, or if the exponent $\tilde{\alpha}$ is near unity, more are required.

ADDITIONAL COMMENTS

Although output need not be produced, we recommend calling STABLE with IUNIT>0, so the user will have a record of how estimation progressed. The following information will then be printed out:

by MONIT: number of E04KBF iterations and function evaluations, objective function value, norm of projected gradient, subproblem solution, projected gradient, and estimated condition number of projected Hessian;

by STABLE: E04KBF fault indicator, and value of convergence criterion;

by HESDIF(if called): number of iterations needed to compute approximate Hessian, and steplength used.

Use of STABLE in "batch mode" has drawbacks. For instance, most faults arising in E04KBF are not diagnosed until iteration ceases. In practice, such faults may likely be due to the initial $\tilde{\alpha}$ being on the wrong side of 1. Further, when $\tilde{\alpha}$ is small, convergence tolerances are difficult to interpret, and the user may prefer direct control of iteration. We therefore prefer to use STABLE interactively, a copy of the output described above being directed to the terminal, and the user deciding after each iteration whether he wishes to continue. Required modifications are simple.

Faster and/or more compact codings of the Algorithm are possible, for instance, if β is known to be zero, if only Formulation A or Formulation B is desired, or if asymptotic covariance matrices are not desired. Generality is achieved at a price in efficiency.

REFERENCES

- Brockwell, P.J. and Brown, B.M. (1981). High-efficiency estimation for the positive stable laws. J. Amer. Statist. Assoc., 76, 626-631.
- Chambers, J.M. (1977). Computational Methods for Data Analysis. New York: Wiley.
- Chambers, J.M., Mallows, C.L. and Stuck, B.W. (1976). A method for simulating stable random variables. J. Amer. Statist. Assoc., 71, 340-344.
- DuMouchel, W.H. (1971). Stable distributions in statistical inference. Unpublished Ph.D. thesis, Department of Statistics, Yale University.
- Fama, E.F. and Roll, R. (1971). Parameter estimation for symmetric stable distributions. J. Amer. Statist. Assoc., 66, 331-338.
- Feuerverger, A. and McDunnough, P. (1981a). On the efficiency of empirical characteristic function procedures. J. R. Statist. Soc. B, 43, 20-27.
- Feuerverger, A. and McDunnough, P. (1981b). On some Fourier methods for inference. J. Amer. Statist. Assoc., 76, 379-387.
- Hiebert, K.L. (1981). An evaluation of mathematical software that solves nonlinear least squares problems. ACM Transactions on Mathematical Software, 7, 1-16.
- Koutrouvelis, I.A. (1980). Regression-type estimation of the parameters of stable laws. J. Amer. Statist. Assoc., 75, 918-928.
- Koutrouvelis, I.A. (1981). An iterative procedure for the estimation of the parameters of stable laws. Communications in Statistics B, 10, 17-28.
- Mandelbrot, B. (1963). The variation of certain speculative prices. Journal of Business, 36, 394-419.
- NAG FORTRAN Library Manual, Mark 8, The Numerical Algorithms Group (USA) Inc., Downers Grove, ILL., 1981.
- Paulson, A.S., Holcomb, E.W. and Leitch, R.A. (1975). The estimation of the parameters of the stable laws. Biometrika, 62, 163-170.
- Paulson, A.S., Delehanty, T.A. and Brothers, K.M. (1982). Some properties of modified integrated squared error estimators for the stable laws. Submitted for publication.
- Paulson, A.S., and Delehanty, T.A. (1982). Modified weighted squared error estimation procedures with special emphasis on the stable laws. Submitted for publication.

```

C*****STAB 001
C CENTRAL SUBROUTINE OF ESTIMATION PROCESS.
C CALLED BY - MAIN
C CALLS - GRIDMT, SETEGF, VARIAB
C PASSES TO E04KBF VIA EXTERNAL SYNT - FUNCT, MONIT, E04LBS
C M.A.G. PROCEDURES CALLED -
C X02AAF (RETURNS MACHINE PRECISION),
C E04KBF (QUASI-NEWTON OPTIMIZATION WITH BOUNDED
C E04KBF VARIABLES),
C F01CAF (SETS A MATRIX TO ZERO).
C*****STAB 002
C SUBROUTINE STABLE(X, N, MODE, KSUM, XLAM, TAU, NPTS, TOL, MAXIT,
C *XL, XB, XH, NPAR, ISCLBD, LOCBND, ICOV, VCV1, VCV2, WORK, LWORK,
C *LWORK, LIWORK, IUNIT, IFAULT)
C*****STAB 003
C EXTERNAL ROUTINES
C EXTERNAL E04LBS, FUNCT, MONIT
C ARGUMENTS
C INTEGER N, MODE, KSUM, NPTS, MAXIT, NPAR, ISCLBD, LOCBND, ICOV,
C *LWORK, LIWORK, IWORK(LIWORK), IUNIT, IFAULT
C REAL X(N), XLAM, TAU, TOL, XL(NPAR), XH(NPAR),
C *VCV1(NPAR, NPAR), VCV2(NPAR, NPAR), WORK(LWORK)
C FUNCTION CALLED
C REAL X02AAF
C*****STAB 004
C LOCAL SCALARS
C SPECIFICALLY FOR E04KBF PARAMETERS -
C LOGICAL LOGSCH
C INTEGER LW, LIW, MAXCAL, IBOUND, IPRINT, LHESL, INTYPE
C REAL ETA, XTOL, STEPX, FST
C*****STAB 005
C SCRATCH VARIABLES, SUBSCRIPT INDICATORS, AND CONSTANTS -
C INTEGER IND, IND1, IOVFLW, NPTS2, IA, NVAR, NVART, NPTS, MMT,
C *MEGF, MDERIV, MWORK, MA, MAINV, MHESL, MHESD, MIW, MW, MV
C REAL SIGMA, XMU, XK, ZERO, PT2, TWOVPI, ONE, TWO, SORT10, FIVE,
C *TEN
C*****STAB 006
C DATA LOGSCH, MAXCAL, IBOUND, IPRINT, ETA, STEPX, FST
C */.TRUE., 400, 0, 0, 0.9, 0.25, 0.0/
C DATA ZERO, PT2, TWOVPI, ONE, TWO, SORT10, FIVE, TEN
C */0.0, 0.2, 0.6366197724, 1.0, 2.0, 3.162277660, 5.0, 10.0/
C*****STAB 007
C ON INPUT ERRORS (IFAU = 1 - 9), EXIT IMMEDIATELY.
C IWORK(1) = 0
C WORK(1) = ZERO
C IFAULT = 1
C IF (MAXIT .LE. 0) RETURN
C IFAULT = 2
C IF (N .LT. 50) RETURN
C IFAULT = 3
C IF (KSUM .LE. 0) RETURN
C IFAULT = 4
C IF (TAU .LT. ZERO .OR. MODE .NE. 0 .AND. TAU .GT. ONE) RETURN
C IFAULT = 5
C IF (NPTS .LT. 20 .OR. MOD(NPTS,2) .NE. 0) RETURN
C IFAULT = 6
C IF (TOL .LE. ZERO) RETURN
C IFAULT = 7
C IF (NPAR .NE. 4) RETURN
C IFAULT = 8
C NPTS2 = NPTS / 2
C LHESL = (NPAR*NPAR - NPAR) / 2
C LW = 10 + 11 * NPAR + LHESL + (3 + NPAR) * NPTS + NPTS2
C IF (MODE .EQ. 0) LW = LW + NPTS * (2*NPTS + 1)
C*****STAB 008
C*****STAB 009
C*****STAB 010
C*****STAB 011
C*****STAB 012
C*****STAB 013
C*****STAB 014
C*****STAB 015
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C*****STAB 059
C*****STAB 060

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C
C
IF (LWORK .LT. LW .OR. LIWORK .LT. NPAR + 7) RETURN
C
C      CHECKING OF PARAMETER BOUNDS (IF FAULT = 9 IF WRONG)
C      IF FAULT = 9
C      IF (XL(1) .GT. XB(1) .OR. XL(1) .GT. XH(1) .OR. XB(1) .GT. XH(1))
C      *RETURN
C      IF (XL(1) .LE. ZERO .OR. XH(1) .GT. TWO) RETURN
C      IF (XL(2) .GT. XB(2) .OR. XL(2) .GT. XH(2) .OR. XB(2) .GT. XH(2))
C      *RETURN
C      IF (XL(2) .LT. - ONE .OR. XH(2) .GT. ONE) RETURN
C      IF ((XL(2) .NE. ZERO .OR. XH(2) .NE. ZERO) .AND. (XL(1) .LT. ONE
C      *AND. XH(1) .GE. ONE .OR. XL(1) .LE. ONE .AND. XH(1) .GT. ONE))
C      *RETURN
C      IF (XB(3) .LE. ZERO) RETURN
C      IF (ISCLBD .NE. 0 .AND. (XL(3) .GT. XB(3) .OR. XL(3) .GT. XH(3) .
C      *OR. XB(3) .GT. XH(3))) RETURN
C      IF (ISCLBD .LT. 0 .AND. XL(3) .NE. XH(3)) RETURN
C      IF (LOCBND .LT. 0 .AND. (XL(4) .GT. XB(4) .OR. XL(4) .GT. XH(4) .
C      *OR. XB(4) .GT. XH(4))) RETURN
C      IF (LOCBND .LT. 0 .AND. XL(4) .NE. XH(4)) RETURN
C      IF (LOCBND .NE. 0 .AND. XL(1) .EQ. ONE .AND. XH(1) .EQ. ONE .AND.
C      *(XL(2) .NE. ZERO .OR. XL(2) .NE. XH(2))) RETURN
C      *
C      INITIAL ADJUSTMENT OF LOCATION/SCALE PARAMETERS/BOUNDS.
C      SAVE BOUNDS FOR EXIT.
C      WORK(NPAR + 5) = XL(3)
C      WORK(NPAR + 6) = XH(3)
C      WORK(NPAR + 7) = XL(4)
C      WORK(NPAR + 8) = XH(4)
C      XK = FLOAT(KSUM)
C      LOCATION
C      XNW = XK * XB(4)
C      IF (LOCBND .LT. 0) GO TO 10
C      IF (LOCBND .GT. 0) GO TO 20
C      XL(4) = -FIVE
C      XH(4) = FIVE
C      GO TO 30
C      10 XL(4) = ZERO
C      XH(4) = ZERO
C      GO TO 30
C      20 XL(4) = XK * XL(4)
C      XH(4) = XK * XH(4)
C      SCALE
C      30 XTOL = XK ** (ONE/XB(1))
C      SIGMA = XTOL * XB(3)
C      IF (ISCLBD .LT. 0) GO TO 40
C      IF (ISCLBD .GT. 0) GO TO 50
C      XL(3) = PT2
C      XH(3) = FIVE
C      GO TO 60
C      40 XL(3) = ONE
C      XH(3) = ONE
C      GO TO 60
C      50 XL(3) = XTOL * XL(3)
C      XH(3) = XTOL * XH(3)
C      *
C      EXPLICITLY SET XTOL TO E04KBF DEFAULT VALUE, USING X02AAF,
C      SO THAT IT IS AVAILABLE ON EXIT ON E04KBF.
C      60 XTOL = TEN * Sqrt(X02AAF(XTOL))
C
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C
MEMORY MANAGEMENT
MIV = 2 + NPAR
IWORK(MIV + 2) = MODE
IWORK(MIV + 3) = IUNIT
IWORK(MIV + 4) = NPTS
IWORK(MIV + 5) = 9 * NPAR + 2
LIW = 6
LV = LW - 8 - 2 * NPAR - LHESL
MHESL = 9 + NPAR
MHESD = MHESL + LHESL
MW = MHESD + NPAR
NPTS = MW + 9 * NPAR + 2
MWT = NPTS + NPTS2
MECF = MWT + NPTS
MDERIV = MECF + NPTS
      AVOID UNCLEAR REFERENCES TO A AND A INVERSE IF MODE .NE. 0
MA = MDERIV
MAINV = MDERIV
IA = NPTS + 1
IF (MODE .NE. 0) GO TO 70
MA = MA + (NPAR + 1) * NPTS
MAINV = MA + IA * NPTS
      LOOPING POINT FOR ITERATION
70 IWORK(1) = IWORK(1) + 1
IOVFLW = 0

C
      IF LOCATION/SCALE PARAMETERS ARE CONSTRAINED, UPDATE
      THEIR BOUNDS.
      IF (ISCLBD .LE. 0) GO TO 80
      XL(3) = AMIN1(ONE, AMAX1(P12, WORK(NPAR + 5)/SIGMA))
      XH(3) = AMAX1(ONE, AMIN1(FIVE, WORK(NPAR + 6)/SIGMA))
80 IF (LOCBND .LE. 0) GO TO 90
      XL(4) = AMIN1(ZERO, AMAX1((-FIVE, (WORK(NPAR + 7) - XMU)/SIGMA))
      XH(4) = AMAX1(ZERO, AMIN1(FIVE, (WORK(NPAR + 8) - XMU)/SIGMA))
      CHANGE PARAMETERS TO REFLECT FUTURE STANDARDIZATION.
90 XB(3) = ONE
      XB(4) = ZERO
      WORK(NPAR + 3) = XB(1)
      WORK(NPAR + 4) = XB(2)
      SET INTYPE = 3 IF POSSIBLE, SO OLD HESSIAN CAN BE USED.
      INTYPE = 0
      IF (IWORK(1) .EQ. 1) GO TO 110
      INTYPE = 3
      DO 100 I = 1, NPAR
      IF (IWORK(I + 1) .GT. 0) GO TO 100
      INTYPE = 0
      GO TO 110
100 CONTINUE

C
      SET GRIDPOINTS, WEIGHTS.
110 CALL GRIDWT(XB, NPAR, XLAM, TAU, WORK(NPTS), NPTS2, WORK(MWT),
      *NPTS, MODE, KSUM)
C
      SET E. CH. F. VALUES, ALSO A AND A INVERSE, IF MODE = 0, USING
      FIRST COL. OF DERIV AS WORKSPACE.
      CALL SETCF(X, N, XB, NPAR, MODE, TAU, SIGMA, XMU, KSUM, IA,
      *NPTS2, NPTS, WORK(NPTS), WORK(MECF), WORK(MA), WORK(MAINV),
      *WORK(MDERIV), IFAULT)
      IF (IFault .GT. 0) IFAULT = 9 + IFAULT
      IF (IFault .EQ. 10 .OR. IFAULT .EQ. 11) GO TO 180
C

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C      CALL OR RESTART M.A.G. ROUTINE E04KBF
120  IFALT = 1
    CALL E04KBF(NPAR, FUNCT, MONIT, IPRINT, LOCSCN, INTYPE, E04LBS,
    *MAXCAL, ETA, XTOL, STEPMX, FST, IBOUND, XL, XH, XB, WORK(MHESL),
    *LHESL, WORK(MHESD), IWORK(2), WORK(2), WORK(3), IWORK(MIW), LIW,
    *WORK(MW), LW, IFAULT)
C      COMPUTE CONVERGENCE CRITERION
    WORK(1) = AMAX1(ABS(XB(3) - ONE), ABS(XB(4)))
    IF (ISCLBD .LT. 0 .AND. LOCEND .LT. 0) WORK(1) = AMAX1(ABS(XB(1) -
    *WORK(NPAR + 3)), ABS(XB(2) - WORK(NPAR + 4)))
C      OUTPUT IF REQUESTED
    IF (IUNIT .GT. 0) WRITE (IUNIT, 1000) IWORK(1), IFAULT, WORK(1)
    IF (IFAULT .EQ. 0) GO TO 140
    IF E04KBF FAULT INDICATOR IS 2, 3, 4, OR 5, JUDGE QUALITY
C      OF SOLUTION. NOTE IFAULT = 1 IS IMPOSSIBLE.
    IFAULT = 10 + IFAULT
    IF MAXCAL FUNCTION EVALUATIONS HAVE BEEN MADE, GIVE UP
C      IF (IFAULT .EQ. 12) GO TO 180
    IF (IFAULT .NE. 14) GO TO 130
C      ONE OVERFLOW IN HESSIAN CHOLESKY FACTORS IS ALLOWED
    IF (IOVFLW .GE. 1) GO TO 180
    IOVFLW = IOVFLW + 1
    INTYPE = 0
    GO TO 120
C      IF IFAULT = 13 OR 15, TEST PROJECTED GRADIENT AND PROJECTED
C      HESSIAN CONDITION NUMBER
130  IND = MW + 9 * NPAR
    IF (SQRT10*WORK(IND + 1) .LT. XTOL .AND. WORK(IND) .LT. ONE/WORK(
    *IND + 1)) IFAULT = 0
C      UPDATE LOCATION AND SCALE AND TEST CONVERGENCE
140  XMU = XMU + SIGMA * XB(4)
    IF (XB(1) .EQ. ONE) XMU = XMU + SIGMA * TWOVPI * XB(2) * XB(3) *
    *ALOG(XB(3))
    SIGMA = SIGMA * XB(3)
    IF (WORK(1) .LT. TOL) GO TO 150
    IF (IWORK(1) .LT. MAXIT) GO TO 70
C      MAXIT ITNS HAVE BEEN USED WITHOUT CONVERGENCE - SET IFAULT.
C      IND = 16
    IF (IFAULT .EQ. 13) IND = 17
    IF (IFAULT .EQ. 15) IND = 18
    IFAULT = IND
C      TERMINATION WITH IFAULT = 0, 13, 15, 16, 17 OR 18
150  IF (KSUM .GT. 1 .OR. ICOV .LT. 0) GO TO 190
C      PREPARE TO CALL VARIAB TO COMPUTE COVARIANCE MATRICES.
C      MEMORY MANAGEMENT.
C      MVWORK = MDERIV + NPAR * NPTS
C      MV = MW + NPAR * (NPAR + 1)
C      REARRANGE THE M.A.G. ISTATE VECTOR SO IT HOLDS FREE
C      PARAMETER SUBSCRIPTS IN INCREASING ORDER.
C      NVAR = 0
DO 160 I = 1, NPAR
    IF (IWORK(I + 1) .LT. 0) GO TO 160
    NVAR = NVAR + 1
    IWORK(NVAR + 1) = I
160  CONTINUE

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C      IN THE RARE EVENT THAT THERE ARE NO FREE VARIABLES, VARIAB
C      IS NOT CALLED AND COVARIANCE MATRICES ARE SET TO ZERO.
      IF (NVAR .GT. 0) GO TO 170
      IF (ICOV .EQ. 0) CALL F01CAF(VCV1, NPAR, NPAR, IWORK(2))
      CALL F01CAF(VCV2, NPAR, NPAR, IWORK(3))
      GO TO 190
170  NVAR1 = NVAR + 1
      XB(3) = ONE
      XB(4) = ZERO

C      COMPUTE ESTIMATED ASYMPTOTIC COVARIANCE MATRICES.
C      CALL VARIAB(ICOV, X, N, XB, NPAR, MODE, SIGMA, XMU, IWORK(2),
      *NVAR, WORK(MPTS), NPTS2, WORK(MMT), WORK(MECF), NPTS,
      *WORK(MDERIV), WORK(MVWORK), WORK(MHESD), WORK(MA), IA,
      *WORK(MAINV), VCV1, VCV2, WORK(MM), NVAR1, WORK(MV), IWORK(MIW),
      *LIW, WORK(MM), LW, IND, IND1)
      SAVE FAULT INDICATORS AND NO. OF ITNS TAKEN FOR HESSIAN.
      IWORK(2) = IND
      IWORK(3) = IND1
      IWORK(4) = IWORK(MIW)
      GO TO 190

C      EXIT FOR IFAULT = 10, 11, 12, OR 14. IF IFAULT = 12 OR 14,
C      RESULTS OF THE LAST (ABORTED) OPTIMIZATION ARE NOT USED.
180  IWORK(2) = IFAULT
      IWORK(3) = IFAULT

C      EXIT FOR IFAULT = 0, 13, 15, 16, 17, OR 18.
C      RESET LOCATION/SCALE BOUNDS
190  XL(3) = WORK(NPAR + 5)
      XH(3) = WORK(NPAR + 6)
      XL(4) = WORK(NPAR + 7)
      XH(4) = WORK(NPAR + 8)

C      ADJUST PARAMETERS TO STANDARD FORM
      IF (XB(1) .EQ. ONE) XMU = XMU - TWOVPI * XB(2) * SIGMA * ALOG(
      *SIGMA)
      XB(4) = XMU / XK
      XB(3) = SIGMA / (XK**(ONE/XB(1)))

C      1000 FORMAT (10HMAJOR ITN, I3, 18H, IFAIL (E04KBF) =, I3,
      *25H, CONVERGENCE CRITERION =, E11.3)
      RETURN
      END

C*****
C      CALCULATES (POSITIVE) GRIDPOINTS AND ALL WEIGHTS.
C      CALLED BY - STABLE
C      CALLS - CHARFM
C*****
C      SUBROUTINE GRIDNT(PAR, NPAR, XLAM, TAU, PTS, NPTS2, WT, NPTS,
      *MODE, KSUM)
C      ARGUMENTS
C      INTEGER NPAR, NPTS2, NPTS, MODE, KSUM
C      REAL PAR(NPAR), XLAM, TAU, PTS(NPTS2), WT(NPTS)
C      LOCAL SCALARS
C      LOGICAL FLAG
C      INTEGER IND, IND1, IND2, INNER
C      REAL ALPHA, AFAC1, AFAC2, TEMP, GAP, END, C1
C      CONSTANTS
C      REAL ZERO, PT025, PT035, PT04, PT0425, PT045, PT05, PT06, PT07,
      *PT075, PT3, PT5, PT6, PT8, PT9, ONE, ONEPT2, ONEPT4, ONEPT6,

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GRID 016
GRID 017

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C *ONEP7, ONEPT8, TWO, THREE, THRPT3, THRPT6, FOUR, FIVE, SEVEN, TENGRID 019
C DATA ZERO, PT025, PT035, PT04, PT0425, PT045, PT05, PT06, PT07, GRID 019
C *PT075, PT3, PT5, PT6, PT8, PT9, ONE, ONEPT2, ONEPT4, ONEPT6, GRID 020
C *ONEP7, ONEPT8, TWO, THREE, THRPT3, THRPT6, FOUR, FIVE, SEVEN, GRID 021
C *TEN /0.0, 0.025, 0.035, 0.04, 0.0425, 0.045, 0.05, 0.06, 0.07, GRID 022
C *.075, 0.3, 0.5, 0.6, 0.8, 0.9, 1.0, 1.2, 1.4, 1.6, 1.7, 1.8, 2.0, GRID 023
C *.3, 0.3, 3.6, 4.0, 5.0, 7.0, 10.0/ GRID 024
C GRID 025
C SELECT NUMBER OF INNER GRIDPOINTS. GRID 026
C NUMBER OF INNER PTS = 2 IF NPTS .LT. 30, 3 IF NPTS .GE. 30. GRID 027
C INNER = 2 GRID 028
C IF (NPTS .GE. 30) INNER = 3 GRID 029
C ALPHA = PAR(1) GRID 030
C IF (ALPHA .LE. ONE) GO TO 70 GRID 031
C CASE WHEN ALPHA .GT. 1 - FIRST CHOOSE RIGHT ENDPOINT. GRID 032
C END = THREE GRID 033
C IF (ALPHA .LT. ONEPT8) END = THRPT3 GRID 034
C IF (ALPHA .LT. ONEPT2) END = THRPT6 GRID 035
C IF (KSUM .EQ. 1) GO TO 20 GRID 036
C GRID 037
C GRID 038
C GRID 039
C GRID 040
C GRID 041
C GRID 042
C GRID 043
C GRID 044
C GRID 045
C GRID 046
C GRID 047
C WHEN KSUM .GT. 1 - FIRST USE HALF OF F-M ALPHA OPTIMAL GAPS GRID 048
C FOR POINTS CLOSE TO ORIGIN. GRID 049
C 20 IF (NPTS .GE. 30) GO TO 30 GRID 050
C GAP = PT075 GRID 051
C IF (ALPHA .LT. ONEPT8) GAP = PT06 GRID 052
C IF (ALPHA .LT. ONEPT6) GAP = PT07 GRID 053
C IF (ALPHA .LT. ONEPT4) GAP = PT075 GRID 054
C IF (ALPHA .LT. ONEPT2) GAP = PT0425 GRID 055
C GO TO 40 GRID 056
C 30 GAP = PT05 GRID 057
C IF (ALPHA .LT. ONEPT8) GAP = PT035 GRID 058
C IF (ALPHA .LT. ONEPT6) GAP = PT04 GRID 059
C IF (ALPHA .LT. ONEPT4) GAP = PT045 GRID 060
C IF (ALPHA .LT. ONEPT2) GAP = PT025 GRID 061
C DO 50 I = 1, INNER GRID 062
C TEMP = TEMP + GAP GRID 063
C PIS(I) = TEMP GRID 064
C 50 CONTINUE GRID 065
C LOGARITHMICALLY SPACE THE REST OF THE POINTS BY LOG(1 + U / 2) GRID 066
C TEMP = ALOG(ONE + TEMP/TWO) GRID 067
C GAP = (ALOG(ONE + END/TWO) - TEMP) / FLOAT(NPTS2 - INNER) GRID 068
C IND = INNER + 1 GRID 069
C DO 60 I = IND, NPTS2 GRID 070
C TEMP = TEMP + GAP GRID 071
C PIS(I) = TWO * (EXP(TEMP) - ONE) GRID 072
C 60 CONTINUE GRID 073
C GO TO 100 GRID 074
C GRID 075
C GRID 076
C GRID 077

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C          CASE WHEN ALPHA .LE. 1 - SUBTRACT 1 FROM NO. OF INNER PTS
C          IF ALPHA .LE. 1/2.  START BY EQUISPACING ALL POINTS FOR
C          LOG(U + (MAX(ALPHA,0.3)**3)), BUT MULTIPLY INNER POINTS BY
C          (MAX(ALPHA,0.3)**2 / 4).  (THE LATTER MULTIPLICATION IS
C          OMITTED IF KSUM .GT. 1.)  THEN CONTINUE SPACING.
70  END = FIVE
   IF (ALPHA .LT. ONE) END = FOUR
   IF (ALPHA .LT. PT9) END = FIVE
   IF (ALPHA .LT. PT8) END = SEVEN
   IF (ALPHA .LT. PT6) END = TEN
   AFAC1 = AMAX1(ALPHA,PT3)
   AFAC2 = ONE
   IF (KSUM .EQ. 1) AFAC2 = AFAC1 * AFAC1 / FOUR
   AFAC1 = AFAC1 * AFAC1 * AFAC1
   TEMP = ALOG(AFAC1)
   C1 = ALOG(END + AFAC1)
   GAP = (C1 - TEMP) / FLOAT(NPTS2)
   IF (ALPHA .LE. PT5) INNER = INNER - 1
   DO 80 I = 1, INNER
     TEMP = TEMP + GAP
     PTS(I) = (EXP(TEMP) - AFAC1) * AFAC2
80  CONTINUE
     TEMP = ALOG(PTS(INNER) + AFAC1)
     GAP = (C1 - TEMP) / FLOAT(NPTS2 - INNER)
     IND = INNER + 1
     DO 90 I = IND, NPTS2
       TEMP = TEMP + GAP
       PTS(I) = EXP(TEMP) - AFAC1
90  CONTINUE

C          COMPUTE WEIGHTS (ALL NPTS OF THEM).
C          IF KSUM .GT. 1, ALL WEIGHTS ARE 1.
100 IF (KSUM .EQ. 1) GO TO 120
   DO 110 I = 1, NPTS
     WT(I) = ONE
   RETURN

C          FLAG = MODE .NE. 0
   IND = NPTS2 + 1
   DO 140 I = 1, NPTS2
     TEMP = PTS(I)
     IND1 = NPTS2 + 1
     IND2 = IND - 1
     GAP = EXP(-XLAM*TEMP**ALPHA)
     IF (FLAG) GO TO 130
     WT(IND1) = GAP
     WT(IND2) = GAP
     GO TO 140
130 GAP = GAP * GAP
     CALL CHARFN(TEMP, PAR, NPAR, END, AFAC1)
     CALL CHARFN(TEMP + TEMP, PAR, NPAR, AFAC2, C1)
     WT(IND1) = GAP / (ONE + TAU*(C1 - (END + AFAC1)**2))
     WT(IND2) = GAP / (ONE - TAU*(C1 + (END - AFAC1)**2))
140 CONTINUE

C          RETURN
   END
C*****
C          COMPUTES REAL AND IMAGINARY PARTS OF STANDARD STABLE
C          CHARACTERISTIC FUNCTION (SIGMA = 1, MU = 0).

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GRID 134
CHAR 001
CHAR 002
CHAR 003

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C          CALLED BY - GRIDINT, SEIECF, VARIAB          CHAR 004
C*****          *****          *****          CHAR 005
C          SUBROUTINE CHARFNI(U, PAR, NPAR, RE, XIM)          CHAR 006
C          ARGUMENTS          CHAR 007
C          INTEGER NPAR          CHAR 008
C          REAL U, PAR(NPAR), RE, XIM          CHAR 009
C          LOCAL SCALARS          CHAR 010
C          REAL ALPHA, XMOD, ZERO, ONE, PIBY2          CHAR 011
C          DATA ZERO, ONE, PIBY2 /0.0, 1.0, 1.570796327/          CHAR 012
C          CHAR 013
C          RE = ABS(U)          CHAR 014
C          ALPHA = PAR(1)          CHAR 015
C          IF (ALPHA .NE. ONE) GO TO 10          CHAR 016
C          XIM = ZERO          CHAR 017
C          IF (U .NE. ZERO) XIM = ALOG(RE) / PIBY2          CHAR 018
C          GO TO 20          CHAR 019
C          10 XIM = TAN(PIBY2*ALPHA)          CHAR 020
C          20 XMOD = RE ** ALPHA          CHAR 021
C          XIM = -PAR(2) * SIGN(XMOD, U) * XIM          CHAR 022
C          XMOD = EXP(-XMOD)          CHAR 023
C          RE = XMOD * COS(XIM)          CHAR 024
C          XIM = XMOD * SIN(XIM)          CHAR 025
C          CHAR 026
C          RETURN          CHAR 027
C          END          CHAR 028
C*****          *****          *****          FUNC 001
C          FUNCTION/DERIVATIVE EVALUATION          FUNC 002
C          CALLED BY - E04KBF, VARIAB, HESDIF          FUNC 003
C*****          *****          *****          FUNC 004
C          SUBROUTINE FUNCT(IFLAG, N, XC, FC, GC, IW, LIW, W, LW)          FUNC 005
C          ARGUMENTS          FUNC 006
C          INTEGER IFLAG, N, LIW, IW(LIW), LW          FUNC 007
C          REAL XC(N), FC, GC(N), W(LW)          FUNC 008
C          LOCAL SCALARS          FUNC 009
C          LOGICAL ALF1, LFLAG          FUNC 010
C          INTEGER ILOW, NPTS, NPTS2, IND, IND1, ISUB, ISUB1, IPTS, IWT,          FUNC 011
C          *IECF, IDERIV, IPSI          FUNC 012
C          REAL ALPHA, BSIGN, SIGMA, XMU, OMEGA, XMOD, PTS1, SINE, COSINE,          FUNC 013
C          *PISEC2, SUEXP, XLOGSU, FAC, Z, Z1, ZERO, ONE, PIBY2          FUNC 014
C          DATA ZERO, ONE, PIBY2 /0.0, 1.0, 1.570796327/          FUNC 015
C          CHAR 016
C          ALPHA = XC(1)          FUNC 017
C          BSIGN = XC(2)          FUNC 018
C          SIGMA = XC(3)          FUNC 019
C          XMU = XC(4)          FUNC 020
C          ALF1 = ALPHA .EQ. ONE          FUNC 021
C          LFLAG = IFLAG .EQ. 0          FUNC 022
C          CHAR 023
C          VARIABLES TO AID ADDRESSING IN W VECTOR - IPTS = ONE LESS THAN          FUNC 024
C          POSITION OF FIRST POSITIVE GRIDPOINT IN W VECTOR, ETC.          FUNC 025
C          NPTS = IW(5)          FUNC 026
C          NPTS2 = NPTS / 2          FUNC 027
C          IND = NPTS + 1          FUNC 028
C          ILOW = NPTS2 + 1          FUNC 029
C          IPTS = IW(6)          FUNC 030
C          IWT = IPTS + NPTS2          FUNC 031
C          IECF = IWT + NPTS          FUNC 032
C          IDERIV = IECF + NPTS          FUNC 033
C          IPSI = IDERIV + N * NPTS          FUNC 034
C          NPTS2 = IPTS - NPTS2          FUNC 035

```

```

C      IF ALPHA.NE.1, COMPUTE OMEGA(U,ALPHA) AND ITS DERIVATIVE W. R.
C      TO ALPHA OUTSIDE MAIN LOOP.
C      IF (ALF1) GO TO 10
C      OMEGA = TAN(PIBY2*ALPHA)
C      PISEC2 = PIY2 * (ONE + OMEGA*OMEGA)
C      LOOP OVER POSITIVE GRID PTS (SGN(U) IGNORED), SAVE PSI AND
C      ITS GRADIENT AT ALL GRID PTS (PSI(U) = RE(PHI(U)) +
C      IM(PHI(U)) - EMPIRICAL COUNTERPARTS.)
C      DO 20 I = ILOW, NPTS
C      IND1 = IND - I
C      ISUB = NPTS2 + I
C      PTSJ = W(ISUB)
C      XLOGSU = SIGMA * PTSI
C      SUEXP = XLOGSU ** ALPHA
C      XLOGSU = ALOG(XLOGSU)
C      IF (ALF1) OMEGA = XLOGSU / PIY2
C      COSINE = XMU * PTSI - SUEXP * BSIGN * OMEGA
C      SINE = SIN(COSINE)
C      COSINE = COS(COSINE)
C      XMOD = EXP(-SUEXP)
C      SUEXP = SUEXP * XMOD
C      SAVE COMPONENTS OF PSI
C      ISUB = IPSI + I
C      ISUB1 = IECF + I
C      W(ISUB) = XMOD * (COSINE + SINE) - W(ISUB1)
C      ISUB = IPSI + IND1
C      ISUB1 = IECF + IND1
C      W(ISUB) = XMOD * (COSINE - SINE) - W(ISUB1)
C      IF (LFLAG) GO TO 20
C      CALCULATE DERIVATIVES IF REQUIRED
C      DERIVATIVES W. R. TO ALPHA
C      FAC = XLOGSU * OMEGA
C      IF (.NOT. ALF1) FAC = FAC + PISEC2
C      FAC = FAC * BSIGN
C      Z = FAC + XLOGSU
C      Z1 = FAC - XLOGSU
C      ISUB = IDERIV + I
C      W(ISUB) = SUEXP * (-Z*COSINE + Z1*SINE)
C      ISUB1 = IDERIV + IND1
C      W(ISUB1) = SUEXP * (Z1*COSINE + Z*SINE)
C      DERIVATIVES W. R. TO BETA
C      FAC = SUEXP * OMEGA
C      ISUB = ISUB + NPTS
C      W(ISUB) = FAC * (SINE - COSINE)
C      ISUB1 = ISUB1 + NPTS
C      W(ISUB1) = FAC * (SINE + COSINE)
C      DERIVATIVES W. R. TO SIGMA
C      FAC = BSIGN * OMEGA
C      Z = ONE + FAC
C      Z1 = ONE - FAC
C      FAC = -ALPHA * SUEXP / SIGMA
C      ISUB = ISUB + NPTS
C      W(ISUB) = FAC * (Z*COSINE + Z1*SINE)
C      ISUB1 = ISUB1 + NPTS

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FUNC 036
 FUNC 037
 FUNC 038
 FUNC 039
 FUNC 040
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 FUNC 095

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C
C      W(ISUB1) = FAC * (Z1 * COSINE - Z * SINE)
C      DERIVATIVES W. R. TO MU
C      FAC = PTSI * XMOD
C      ISUB = ISUB + NPTS
C      W(ISUB) = FAC * (-SINE + COSINE)
C      ISUB1 = ISUB1 + NPTS
C      W(ISUB1) = -FAC * (SINE + COSINE)
C      20 CONTINUE
C
C      NOW COMPUTE OBJECTIVE FUNCTION , OPTIONALLY GRADIENT.
C      FC = ZERO
C      IF (LFLAG) GO TO 40
C      ILOW = IDERIV - NPTS
C      DO 30 I = 1, N
C      GC(I) = ZERO
C      30 GC(I) = ZERO
C      40 IF (IW(3) .EQ. 0) GO TO 70
C
C      SUM-OF-SQUARES ESTIMATION
C      DO 60 I = 1, NPTS
C      FM. EVALUATION
C      ISUB = IPSI + I
C      Z = W(ISUB)
C      ISUB = IWT + I
C      Z1 = Z * W(ISUB)
C      FC = FC + Z * Z1
C      IF (LFLAG) GO TO 60
C      GRADIENT EVALUATION IF REQUESTED.
C      ISUB = ILOW + I
C      Z1 = Z1 + Z1
C      DO 50 J = 1, N
C      ISUB = ISUB + NPTS
C      GC(J) = GC(J) + W(ISUB) * Z1
C      50 CONTINUE
C      60 CONTINUE
C      RETURN
C
C      MATRIX ESTIMATION - FIRST MULTIPLY PSI VALUES BY WEIGHTS.
C      DO 80 I = 1, NPTS
C      ISUB = IPSI + I
C      ISUB1 = IWT + I
C      W(ISUB) = W(ISUB) * W(ISUB1)
C      80 CONTINUE
C      POSITION OF A INVERSE IS REQUIRED.
C      IND1 = IPSI + IND * NPTS
C      DO 110 I = 1, NPTS
C      Z = ZERO
C      IND1 = IND1 + NPTS
C      SUM OVER J OF PSI(J) * AINV(I,J)
C      DO 90 J = 1, NPTS
C      ISUB = IPSI + J
C      ISUB1 = IND1 + J
C      Z = Z + W(ISUB) * W(ISUB1)
C      90 CONTINUE
C      MULTIPLY BY PSI(I) AND ADD TO FM. VALUE
C      ISUB = IPSI + I
C      FC = FC + Z * W(ISUB)
C      IF (LFLAG) GO TO 110
C
C      GRADIENT IF REQUESTED - FOR JTH COMPONENT OF GRADIENT ADD

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FUNC 096
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FUNC 098
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FUNC 100
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FUNC 153
FUNC 154
FUNC 155

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C      2*(JTH DERIVATIVE AT GRIDPOINT I)*WT(I)*(RESULT OF DO 90 LOOP)
ISUB = IWT + I
Z = (Z + Z) * W(ISUB)
ISUB = ILOW + I
DO 100 J = 1, N
  ISUB = ISUB + NPTS
  GC(J) = GC(J) + Z * W(ISUB)
100 CONTINUE
110 CONTINUE

C      RETURN
      END

C*****
C      COMPUTES EMPIRICAL CH. F. VALUES, ADJUSTING FOR LOCATION,
C      SCALE, AND K-SUM INDEX. FOR MATRIX ESTIMATION, THE UPPER
C      TRIANGLE OF A MATRIX IS CALCULATED AND INVERTED.
C      CALLED BY - STABLE
C      CALLS - CHARFV
C      N.A.G. SUBROUTINE CALLED -
C      F01ABF (ACCURATE INVERSION OF POSITIVE DEFINITE
C      SYMMETRIC MATRIX).
C*****
C      SUBROUTINE SETECF(X, N, PAR, NPAR, MODE, TAU, SIGMA, XMU, KSUM,
C      *IA, NPTS2, NPTS, PTS, ECF, A, AINV, WORK, IFAULT)
C      ARGUMENTS
C      INTEGER N, NPAR, MODE, KSUM, IA, NPTS2, NPTS, IFAULT
C      REAL XMU, PAR(NPAR), TAU, SIGMA, XMU, PTS(NPTS2), ECF(NPTS),
C      *A(IA,NPTS), AINV(NPTS,NPTS), WORK(NPTS)
C      LOCAL SCALARS
C      INTEGER IND, IND1, IND2, IND3
C      REAL PTS1, PTSJ, RE, XIM, RE1, XIM1, RPI, RMI, RPI1, RMI1, ZERO,
C      *HALF
C      DATA ZERO, HALF /0.0, 0.5/

C      CALCULATION OF E.C.H.F. VALUES
PTSJ = FLOAT(N)
XIM1 = FLOAT(KSUM)
RPI = XIM1 * HALF
IND = NPTS2 + 1
IND1 = NPTS + 1
DO 20 I = IND, NPTS
  IND2 = I - NPTS2
  PTS1 = PTS(IND2) / SIGMA
  RE = ZERO
  XIM = ZERO

  DO 10 J = 1, N
    RPI1 = X(J) * PTSI
    RE = RE + COS(RPI1)
    XIM = XIM + SIN(RPI1)
  10 CONTINUE

  RE = RE / PTSJ
  XIM = XIM / PTSJ
  RMI = (RE*RE + XIM*XIM) ** RPI
  RPI1 = XIM1 * ATAN2(XIM,RE) - XMU * PTSI
  RE = RMI * COS(RPI1)
  XIM = RMI * SIN(RPI1)
  ECF(I) = RE + XIM

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FUNC 156
FUNC 157
FUNC 158
FUNC 159
FUNC 160
FUNC 161
FUNC 162
FUNC 163
FUNC 164
FUNC 165
FUNC 166
FUNC 167
SETE 001
SETE 002
SETE 003
SETE 004
SETE 005
SETE 006
SETE 007
SETE 008
SETE 009
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SETE 047
SETE 048

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C      ECF(IND2) = RE - XIM
C      20 CONTINUE
C
C      SET UPPER TRIANGLE OF A IF REQUESTED. A GENERATED FROM
C      POSITIVE GRIDPOINTS ONLY - ANTIDIAGONAL COMPUTED TWICE.
C      IF (MODE.NE. 0) RETURN
C      IF FIRST FILL WORK WITH (RE + IM) (PHI) TO SAVE EVALS.
C      DO 30 I = IND, NPTS
C      IND2 = I - NPTS2
C      CALL CHARFN(PTS(IND2), PAR, NPAR, RE, XIM)
C      IND2 = IND1 - I
C      WORK(1) = RE + XIM
C      WORK(IND2) = RE - XIM
C      30 CONTINUE
C
C      COMPUTATION OF A.
C      DO 40 I = IND, NPTS
C      IND2 = I - NPTS2
C      PTS1 = PTS(IND2)
C      IND2 = IND1 - I
C      RP1 = WORK(1)
C      RM1 = WORK(IND2)
C      DO 40 J = IND, I
C      IND3 = J - NPTS2
C      PTSJ = PTS(IND3)
C      IND3 = IND1 - J
C      RP11 = WORK(J)
C      RM11 = WORK(IND3)
C      CALL CHARFN(PTS1 + PTSJ, PAR, NPAR, RE, XIM)
C      CALL CHARFN(PTS1 - PTSJ, PAR, NPAR, RE1, XIM1)
C      A(J, I) = RE1 + XIM - RP1 * RP11
C      A(IND2, J) = RE - XIM1 - RM1 * RP11
C      A(IND3, I) = RE + XIM1 - RP1 * RM11
C      A(IND2, IND3) = RE1 - XIM - RM1 * RM11
C      40 CONTINUE
C
C      DIAGONAL OF A ADDITIVELY INFLATED BY TAU TIMES AVERAGE OF
C      DIAGONAL ELEMENTS.
C      IF (TAU.EQ. ZERO) GO TO 70
C      PTS1 = ZERO
C      DO 50 I = 1, NPTS
C      PTS1 = PTS1 + A(I, I)
C      50 PTS1 = TAU * PTS1 / FLOAT(NPTS)
C      DO 60 I = 1, NPTS
C      A(I, I) = A(I, I) + PTS1
C      60 CONTINUE
C
C      INVERT A USING N.A.G. ROUTINE FOIABF
C      70 IFAULT = 1
C      CALL FOIABF(A, IA, NPTS, AINV, NPTS, WORK, IFAULT)
C      EXIT IF A FOUND NON POSITIVE DEFINITE OR ILL-CONDITIONED.
C      IF (IFAULT.GT. 0) RETURN
C      ARRANGE A INVERSE SO IT IS COMPLETELY FILLED.
C      DO 80 I = 1, NPTS
C      DO 80 J = 1, I
C      AINV(J, I) = AINV(I, J)
C      80 CONTINUE
C
C      RETURN
C      END
C*****
C      MONITORING OF EQ4NBF.
C*****MONI 001
C      002

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C ***** CALLED BY - F04KBF ***** MONI 003
C ***** SUBROUTINE MONIT(M, XC, FC, GC, ISTATE, GPJNRM, COND, POSDEF, MONI 004
* NITER, NF, IW, LIW, W, LW) MONI 005
C ***** ARGUMENTS ***** MONI 006
C ***** LOGICAL POSDEF ***** MONI 007
C ***** INTEGER M, ISTATE(N), NITER, NF, LIW, IW(LIW), LW MONI 008
C ***** REAL XC(N), FC, GC(N), GPJNRM, COND, W(LW) MONI 009
C ***** LOCAL SCALAR ***** MONI 010
C ***** INTEGER IUNIT ***** MONI 011
C ***** STORE HESSIAN CONDITION NUMBER AND PROJECTED GRADIENT NORM IN MONI 012
C ***** WORK VECTOR ***** MONI 013
C ***** IUNIT = 9 * N + 1 ***** MONI 014
C ***** W(IUNIT) = COND ***** MONI 015
C ***** W(IUNIT + 1) = GPJNRM ***** MONI 016
C ***** IUNIT = IW(4) ***** MONI 017
C ***** IF (IUNIT .LE. 0) RETURN ***** MONI 018
C ***** WRITE DETAILS OF OPTIMIZATION PROCESS IF IUNIT .GT. 0 ***** MONI 019
C ***** WRITE (IUNIT,10) NITER, NF, FC, GPJNRM ***** MONI 020
C ***** WRITE (IUNIT,20) (XC(I), I=1,N) ***** MONI 021
C ***** WRITE (IUNIT,30) (GC(I), I=1,N) ***** MONI 022
C ***** WRITE (IUNIT,40) COND ***** MONI 023
C ***** 10 FORMAT (5HITNS, 5X, 8HFN EVALS, 8X, 8HFN VALUE, 5X, MONI 024
* 21HNORM OF PROJ GRADIENT/1H, 16, 8X, 15, 5X, E11.4, 15X, E11.4) MONI 025
C ***** 20 FORMAT (10HOSOLUTIONS, 4E13.5) ***** MONI 026
C ***** 30 FORMAT (10H PROJ GRAD, 4E13.5) ***** MONI 027
C ***** 40 FORMAT (50H ESTIMATED CONDITION NUMBER OF PROJECTED HESSIAN =, MONI 028
* E12.2) ***** MONI 029
C ***** RETURN ***** MONI 030
C ***** END ***** MONI 031
C ***** CALCULATES EMPIRICAL (VCV1) AND ASYMPTOTIC (VCV2) ***** VARI 001
C ***** APPROXIMATIONS TO ASYMPTOTIC COVARIANCE MATRICES. IFAIL1 AND VARI 002
C ***** IFAIL2 ARE FAILURE INDICATORS FOR MATRIX INVERSION REQUIRED ***** VARI 003
C ***** FOR VCV1, VCV2 RESPECTIVELY. ***** VARI 004
C ***** CALLED BY - STABLE ***** VARI 005
C ***** CALLS - HESDIF, FUNCT, CHARFN, SETVCV, VMATRX, DAPROD, HVPROD ***** VARI 006
C ***** N.A.G. SUBROUTINES CALLED - ***** VARI 007
C ***** FOICAF (SETS A MATRIX TO ZERO), ***** VARI 008
C ***** FOICMF (SET ONE MATRIX TO ANOTHER). ***** VARI 009
C ***** ***** VARI 010
C ***** SUBROUTINE VARIAB(ICOV, X, N, PAR, NPAR, MODE, SIGMA, XMU, ISUB, ***** VARI 011
* NVAR, PTS, NPTS2, WT, ECF, NPTS, DERIV, WORK, HOLD, A, IA, AINV, ***** VARI 012
* VCV1, VCV2, M, NVAR1, V, IW, LIW, W, LW, IFAIL1, IFAIL2) ***** VARI 013
C ***** ARGUMENTS ***** VARI 014
C ***** INTEGER ICOV, N, NPAR, MODE, NVAR, ISUB(NVAR), NPTS2, NPTS, IA, ***** VARI 015
* NVAR1, LIW, IW(LIW), LW, IFAIL1, IFAIL2 ***** VARI 016
C ***** REAL X(N), PAR(NPAR), SIGMA, XMU, PTS(NPTS2), WT(NPTS), ECF(NPTS), ***** VARI 017
* DERIV(NPTS, NPAR), WORK(NPTS), HOLD(NVAR), A(IA, NPTS), ***** VARI 018
* AINV(NPTS, NPTS), VCV1(NPAR, NPAR), VCV2(NPAR, NPAR), H(NVAR1, NVAR), ***** VARI 019
* V(NVAR, NVAR), W(LW) ***** VARI 020
C ***** LOCAL SCALARS ***** VARI 021
C ***** LOGICAL FLAG ***** VARI 022
C ***** INTEGER IND, IND1, IND2, IND3 ***** VARI 023
C ***** REAL PTSK, PTSL, RPI, RMI, RPI1, RM11, COVKL, COVML, COVKM, COVMM, ***** VARI 024
* D1, D2, D3, D4, ZERO ***** VARI 025
C ***** DATA ZERO /0.0/ ***** VARI 026

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C      IND = NPTS2 + 1
C      IND1 = NPTS + 1
C
C      IF REQUESTED, FIRST CALCULATE HESSIAN FOR EMPIRICAL VERSION
C      AND STORE IN VCV1. (HESDIF, FUNCT REQUIRE ECF, WORK,
C      POSSIBLY AINV, WHICH ARE USED AS WORK AREAS BELOW.)
C      H AND V ARE PASSED AS WORKSPACE.
C      IF (ICOV .EQ. 0) CALL HESDIF(PAR, NPAR, ISUB, VCV1, H, V, NVAR,
C      *IW, LIW, W, LW)
C
C      CALL FUNCT TO SET DERIV, AS GRID SEARCH PERFORMED BY E04KBF
C      MAY HAVE SET IT TO STRANGE VALUES. VCV2 USED AS WORKSPACE.
C      CALL FUNCT(2, NPAR, PAR, D1, VCV2(1,1), IW, LIW, W, LW)
C
C      SHIFT DERIV VALUES SO THEY CAN BE ADDRESSED WITHOUT THE
C      ISUB VECTOR. NOTE ISUB ELEMENTS ARE IN ASCENDING ORDER.
C      DO 20 I = 1, NVAR
C        IND2 = ISUB(I)
C      IF (IND2 .EQ. 1) GO TO 20
C      DO 10 J = 1, NPTS
C        DERIV(J,I) = DERIV(J,IND2)
C      20 CONTINUE
C
C      FILL ECF VECTOR WITH CH. F. VALUES, NOW THAT IT IS NO
C      LONGER NEEDED FOR FUNCTION EVALUATION.
C      DO 30 I = IND, NPTS
C        IND2 = 1 - NPTS2
C      CALL CHARFN(PTS(IND2), PAR, NPAR, D1, D2)
C      IND2 = IND1 - 1
C      ECF(I) = D1 + D2
C      ECF(IND2) = D1 - D2
C      30 CONTINUE
C      IF (MODE .EQ. 0) GO TO 100
C
C      SUM OF SQUARES ESTIMATION SECTION.
C
C      FIRST DO ASYMPTOTIC VERSION.
C      CALCULATE UPPER TRIANGLE OF HESSIAN.
C      DO 50 I = 1, NVAR
C      DO 50 J = 1, NVAR
C      D1 = ZERO
C      DO 40 K = 1, NPTS
C      D1 = D1 + DERIV(K,I) * DERIV(K,J) * WT(K)
C      H(I,J) = D1
C      50 CONTINUE
C
C      PREMULTIPLY DERIV BY WEIGHTS TO SAVE MULTIPLICATIONS.
C      DO 60 I = 1, NPTS
C      D1 = WT(I)
C      DO 60 J = 1, NVAR
C      DERIV(I,J) = D1 * DERIV(I,J)
C      60 CONTINUE
C
C      COMPUTE UPPER TRIANGLE OF V. BASICALLY EQUIVALENT TO
C      CALCULATING THE A MATRIX FOR MATRIX ESTIMATION, BUT
C      COMPLICATIONS ARISE IN COMPUTING BILINEAR FORMS.
C      CALL F01CAF(V, NVAR, NPAR, IFAIL2)
C      DO 90 K = IND, NPTS
C      IND2 = K - NPTS2

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VARI 088
VARI 089
VARI 090
VARI 091
VARI 092
VARI 093
VARI 094
VARI 095
VARI 096
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VARI 146
VARI 147

      PTSK = PTS(IND2)
      IND2 = IND1 - K
      RPI = ECF(K)
      RM1 = ECF(IND2)
      DO 90 L = IND, K
      IND3 = L - NPTS2
      PTSL = PTS(IND3)
      IND3 = IND1 - L
      RPI1 = ECF(L)
      RM11 = ECF(IND3)
      CALL CHARFM(PTSK + PTSL, PAR, NPAR, D1, D2)
      CALL CHARFM(PTSK - PTSL, PAR, NPAR, D3, D4)
      COVKL = D3 + D2 - RPI * RPI1
      COVWL = D1 - D4 - RM1 * RPI1
      COVKM = D1 + D4 - RPI * RM11
      COVMM = D3 - D2 - RM1 * RM11
      FLAG = K.EQ.L
      LOOP TO ADD CONTRIBUTIONS TO V (BILINEAR FORMS).
      DO 80 I = 1, NVAR
      D1 = DERIV(K, I)
      D2 = DERIV(IND2, I)
      DO 80 J = 1, NVAR
      D3 = DERIV(L, J)
      D4 = DERIV(IND3, J)
      PTSL = (COVKL*D1 + COVWL*D2) * D3 + (COVKM*D1 + COVMM*D2) * D4
      IF (FLAG) GO TO 70

C
      WHEN K .NE. L, MUST ADD SYMMETRIC CONTRIBUTION.
      RPI1 = DERIV(K, J)
      RM11 = DERIV(IND2, J)
      D3 = DERIV(L, I)
      D4 = DERIV(IND3, I)
      PTSL = PTSL + (COVKL*RPI1 + COVWL*RM11) * D3 + (COVKM*RPI1 +
      *COVMM*RM11) * D4
      70 V(I, J) = V(I, J) + PTSL
      80 CONTINUE
      90 CONTINUE
      GO TO 140

C
      MATRIX ESTIMATION SECTION.
      FILL A MATRIX COMPLETELY (UPPER TRIANGLE ONLY ON ENTRY),
      MULTIPLY ELEMENTS OF A BY CORRESPONDING WEIGHTS, MULTIPLY
      DERIV VALUES BY WEIGHTS TO SAVE MULTIPLICATIONS LATER.
      100 DO 130 I = 1, NPTS
      D1 = WT(I)
      DO 110 J = 1, NPTS
      A(I, J) = A(I, J) * D1 * WT(J)
      A(J, I) = A(I, J)
      110 CONTINUE
      DO 120 J = 1, NVAR
      120 DERIV(I, J) = DERIV(I, J) * D1
      130 CONTINUE

C
      ASYMPTOTIC VERSION.
      MULTIPLY AINV * DERIV, OVERWRITING CORNER OF AINV AND USING
      HOLD AS WORKSPACE.
      CALL DAPROD(AINV, NPTS, NPTS, DERIV, HOLD, NVAR)
      MULTIPLY A * (AINV CORNER), OVERWRITING CORNER OF A AND

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```

DO 50 I = 1, N
D1 = (X(I) - XMU) / SIGMA
C
C      WORK HOLDS VECTOR OF SINE/COSINE TERMS.
DO 10 J = 1, NPTS
IND2 = J - NPTS2
D2 = PTS(IND2) * D1
D3 = COS(D2)
D2 = SIN(D2)
IND2 = IND1 - J
WORK(J) = ECF(J) - D3 - D2
WORK(IND2) = ECF(IND2) - D3 + D2
IF (FLAG) GO TO 10
C
C      IF MATRIX ESTIMATION, MULTIPLY ELEMENTS OF WORK BY WEIGHTS.
WORK(J) = WORK(J) * WT(J)
WORK(IND2) = WORK(IND2) * WT(IND2)
10 CONTINUE
C
C      MULTIPLY DERIV * WORK, WRITING RESULT TO HOLD. MULTIPLICATION
C      DONE DIRECTLY TO AVOID OVERHEAD OF SUBROUTINE CALL.
DO 30 J = 1, NVAR
D1 = ZERO
DO 20 K = 1, NPTS
20 D1 = D1 + DERIV(K,J) * WORK(K)
HOLD(J) = D1
30 CONTINUE
C
C      ADD HOLD * (HOLD TRANSPOSE) TO V.
DO 40 J = 1, NVAR
D1 = HOLD(J)
DO 40 K = J, NVAR
V(J,K) = V(J,K) + D1 * HOLD(K)
40 CONTINUE
50 CONTINUE
C
C      FINAL CONSTANT FACTOR FOR V.
D1 = FOUR / FLOAT(N)
DO 60 I = 1, NVAR
DO 60 J = 1, NVAR
60 V(I,J) = V(I,J) * D1
C
C      RETURN
C      END
C*****
C      MULTIPLIES THE (NPTS BY NPTS) CORNER OF THE (IFAC1 BY NPTS)
C      MATRIX FAC1 BY THE (NPTS BY NVAR) MATRIX FAC2, OVERWRITING
C      THE UPPER LEFT (NPTS BY NVAR) ELEMENTS OF FAC1.
C      UNFORTUNATELY, N.A.G. ROUTINE F01CKF DOES NOT ALLOW THIS
C      KIND OF OVERWRITING.
C      CALLED BY - VARIAB
C*****
C      SUBROUTINE DAPROD(FAC1, IFAC1, NPTS, FAC2, WORK, NVAR)
C      ARGUMENTS
C      INTEGER IFAC1, NPTS, NVAR
C      REAL FAC1(IFAC1,NPTS), FAC2(NPTS,NVAR), WORK(NVAR)
C      LOCAL SCALARS
C      REAL TEMP, ZERO
C      DATA ZERO /0.0/
C

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 DAPR 016

```

C          USING HOLD AS WORKSPACE.
C          CALL DAPROD(A, IA, NPTS, AINV, HOLD, NVAR)
C
C          MULTIPLY (A CORNER TRANSPOSE) * (AINV CORNER), GIVING
C          UPPER TRIANGLE OF V.
C          CALL HYPROD(A, IA, NVAR, AINV, NPTS, V, NVAR)
C
C          MULTIPLY (AINV CORNER TRANSPOSE) * DERIV, GIVING UPPER
C          TRIANGLE OF HESSIAN.
C          CALL HYPROD(AINV, NPTS, NVAR, DERIV, NPTS, H, NVAR1)
C
C          CODE COMMON TO MATRIX AND SUM OF SQUARES ESTIMATION.
C          COMPUTE ASYMPTOTIC COVARIANCE MATRIX.
C          180 CALL SETVCV(ISUB, NVAR, H, NVAR1, V, HOLD, VCV2, NPAR, SIGMA,
C          *IFAIL2)
C          IF (ICOV .GT. 0) RETURN
C
C          EMPIRICAL VERSION IF REQUESTED.
C          COMPUTE EMPIRICAL V MATRIX, USING WORK AND HOLD AS WORK AREAS.
C          IF (MODE .NE. 0) CALL VMATRX(X, N, MODE, XMU, SIGMA, PTS, NPTS2,
C          *WT, ECF, WORK, NPTS, DERIV, V, HOLD, NVAR)
C          IF (MODE .EQ. 0) CALL VMATRX(X, N, MODE, XMU, SIGMA, PTS, NPTS2,
C          *WT, ECF, WORK, NPTS, AINV, V, HOLD, NVAR)
C
C          RESTORE HESSIAN FROM VCV1, COMPUTE COVARIANCE MATRIX.
C          CALL FOICHF(VCV1, NVAR, H, NVAR1, NVAR, NVAR)
C          CALL SETVCV(ISUB, NVAR, H, NVAR1, V, HOLD, VCV1, NPAR, SIGMA,
C          *IFAIL1)
C
C          RETURN
C          END
C*****
C          CALCULATES EMPIRICAL V MATRIX. IMPLICIT IS COMPUTATION OF
C          EMPIRICAL COVARIANCE KERNEL, BUT IT IS FASTER TO COMPUTE A
C          MATRIX PRODUCT FOR EACH SAMPLE POINT THAN TO CUMULATE THE
C          WHOLE KERNEL. FOR MATRIX ESTIMATION, THE DERIV ARRAY
C          IS ACTUALLY THE UPPER LEFT CORNER OF AINV.
C          CALLED BY - VARIAB
C          N.A.G. SUBROUTINE CALLED -
C          *****
C          SUBROUTINE VMATRX(X, N, MODE, XMU, SIGMA, PTS, NPTS2, WT, ECF,
C          *WORK, NPTS, DERIV, V, HOLD, NVAR)
C          ARGUMENTS
C          INTEGER N, MODE, NPTS2, NPTS, NVAR
C          REAL X(N), XMU, SIGMA, PTS(NPTS2), WT(NPTS), ECF(NPTS),
C          *WORK(NPTS), DERIV(NPTS,NVAR), V(NVAR,NVAR), HOLD(NVAR)
C          LOCAL SCALARS
C          LOGICAL FLAG
C          INTEGER IND, IND1, IND2
C          REAL D1, D2, D3, ZERO, FOUR
C          DATA ZERO, FOUR /0.0, 4.0/
C
C          SET V TO 0.
C          FLAG = MODE .NE. 0
C          IND = NPTS2 + 1
C          IND1 = NPTS + 1
C          CALL FOICAF(V, NVAR, NVAR, IND2)
C
C          MAIN LOOP OVER SAMPLE.

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DO 40 I = 1, NPTS
DO 20 J = 1, NVAR
TEMP = ZERO
DO 10 K = 1, NPTS
10 TEMP = TEMP + FAC1(I,K) * FAC2(K,J)
WORK(J) = TEMP
20 CONTINUE
DO 30 J = 1, NVAR
30 FAC1(I,J) = WORK(J)
40 CONTINUE
C
C RETURN
C END
C*****
C MULTIPLIES THE TRANSPOSED (NPTS BY NVAR) CORNER OF THE
C (IFAC1 BY NVAR) MATRIX FAC1 BY THE (NPTS BY NVAR) MATRIX
C FAC2, GIVING THE UPPER TRIANGLE OF EITHER V OR H.
C CALLED BY - VARIAB
C*****
C SUBROUTINE HVPROD(FAC1, IFAC1, NVAR, FAC2, NPTS, VH, IVH)
C ARGUMENTS
C INTEGER IFAC1, NVAR, NPTS, IVH
C REAL FAC1(IFAC1,NVAR), FAC2(NPTS,NVAR), VH(IVH,NVAR)
C LOCAL SCALARS
C REAL TEMP, ZERO
C DATA ZERO /0.0/
C
DO 20 I = 1, NVAR
DO 20 J = 1, NVAR
TEMP = ZERO
DO 10 K = 1, NPTS
10 TEMP = TEMP + FAC1(K,I) * FAC2(K,J)
VH(I,J) = TEMP
20 CONTINUE
C
C RETURN
C END
C*****
C COMMON OPERATIONS IN COMPUTATION OF COVARIANCE MATRICES.
C INVERTS HESSIAN MATRIX H, CALCULATES (H INVERSE) * V *
C (H INVERSE), OVERWRITING V.
C CALLED BY - VARIAB
C N.A.G. SUBROUTINES CALLED -
C F01ABF (ACCURATE INVERSION OF POSITIVE DEFINITE
C SYMMETRIC MATRIX),
C F01CKF (MATRIX MULTIPLICATION WITH OVERWRITING),
C F01CHF (SET ONE MATRIX EQUAL TO ANOTHER),
C F01CAF (SET A MATRIX TO ZERO).
C*****
C SUBROUTINE SETVCV(ISUB, NVAR, H, NVAR1, V, WORK, VCV, NPAR, SIGMA,
C *IFault)
C ARGUMENTS
C INTEGER NVAR, ISUB(NVAR), NVAR1, NPAR, IFault
C REAL H(NVAR1,NVAR), V(NVAR,NVAR), WORK(NVAR), VCV(NPAR,NPAR),
C *SIGMA
C LOCAL SCALARS
C INTEGER IND, IND1
C REAL TEMP, ZERO
C DATA ZERO /0.0/
C

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C      INVERT H, USING VCV AS WORKSPACE.
      IFALT = 1
      CALL F01ABF(H, NPAR, NPAR, VCV, NPAR, WORK, 1, IFALT)
      IF (1, IFALT, EQ, 0) GO TO 10
      ON FAILURE OF INVERSION, SET VCV TO 0 AND RETURN.
      CALL F01CAF(VCV, NPAR, NPAR, IND)
      RETURN
C      FILL OUT VCV AND V (CURRENTLY ONLY HALF FULL).
      DO 20 I = 1, NPAR
      DO 20 J = 1, I
      V(I,J) = V(J,I)
      VCV(J,I) = VCV(I,J)
      20 CONTINUE
C      SET H TO ITS INVERSE IN SUCH A WAY THAT MULTIPLICATION VIA
      F01CKF WILL BE CORRECT. (NOTE DIMENSIONS IN F01CKF CALL)
      CALL F01CKF(VCV, NPAR, H, NPAR, NPAR, NPAR)
C      MULTIPLY H * V, OVERWRITING V.
      CALL F01CKF(V, H, V, NPAR, NPAR, WORK, NPAR, 3, 1, IFALT)
C      MULTIPLY V * H, OVERWRITING V.
      CALL F01CKF(V, V, H, NPAR, NPAR, WORK, NPAR, 2, 1, IFALT)
C      V NOW CONTAINS COVARIANCE MATRIX FOR FREE VARIABLES. ARRANGE
      ITS CONTENTS IN VCV, ADJUST SIGMA AND MU ENTRIES FOR SCALE.
      CALL F01CAF(VCV, NPAR, NPAR, 1, IFALT)
      DO 30 I = 1, NPAR
      DO 30 J = 1, I
      IND = MIN0(I, SUB(I), 1, SUB(J))
      IND1 = MAX0(I, SUB(I), 1, SUB(J))
      VCV(IND, IND1) = V(I,J)
      30 CONTINUE
      DO 40 J = 3, NPAR
      DO 40 I = 1, J
      TEMP = SIGMA * VCV(I,J)
      IF (1, GE, 3) TEMP = SIGMA * TEMP
      VCV(I,J) = TEMP
      40 CONTINUE
C      FILL LOWER TRIANGLE OF VCV WITH CORRELATIONS.
      DO 50 I = 2, NPAR
      TEMP = VCV(1,1)
      IND = I - 1
      DO 50 J = 1, IND
      IF (AMIN1(TEMP, VCV(J,I)) .LE. ZERO) GO TO 50
      VCV(I,J) = VCV(J,I) / SQRT(TEMP*VCV(J,J))
      50 CONTINUE
C      RETURN
      END
C*****
C      COMPUTES APPROXIMATE UPPER TRIANGLE OF HESSIAN BY DIFFERENCES.
C      NOTE THAT WITH IFLAG = 0, FUNCT WILL NOT SET A GRADIENT.
C      CALLED BY - VARIAB
C      CALLS - FUNCT
C*****
C      SUBROUTINE HESDIF(PAR, NPAR, ISUB, H, SAVE1, SAVE2, NVAR, 1W, LIM, HESD 007
      *W, LW)
C      ARGUMENTS

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HESD 003
HESD 004
HESD 005
HESD 006
HESD 007
HESD 008
HESD 009

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C
INTEGER NPAR, NVAR, ISUB(NVAR), LIW, IW(LIW), LW
REAL PAR(NPAR), H(NVAR,NVAR), SAVE1(NVAR,NVAR), SAVE2(NVAR,NVAR),
  *W(LW)
C
LOCAL SCALARS
INTEGER ITER, IND, IND1, IND2, IND3
REAL PAR1, PARJ, TEMP, TEMP1, STEP, DENOM, ZERO, TOL, TENMU,
  *STEP1, ONE, TWO, SORT10, FOUR
DATA ZERO, TOL, TENMU, STEP1, ONE, TWO, SORT10, FOUR /0.0, 1.0E-6,
  *1.0E-4, 3.162277660E-3, 1.0, 2.0, 3.162277660, 4.0/
C
START WITH STEPLENGTH 1.0E-3, REPEATEDLY DIVIDE BY
SORT10. ITERATION STOPS WHEN DIFFERENCE BETWEEN SUCCESSIVE
HESSIAN LESS THAN 1.0E-6. IF NO SUCCESS IN 5 ITERATIONS,
USE RESULT WITH STEPLENGTH 1.0E-4.
C
DIFFERENCE IS MAX. DIFFERENCE BETWEEN SUCCESSIVE ELEMENTS-
ABSOLUTE DIFFERENCE IF /LATEST ELEMENT/ .LT. 1,
RELATIVE DIFFERENCE IF /LATEST ELEMENT/ .GE. 1.
C
ITER = 0
STEP = STEP1
DO 10 I = 1, NVAR
DO 10 J = 1, NVAR
10 SAVE1(I,J) = ZERO
C
STARTING POINT FOR ITERATION.
20 ITER = ITER + 1
STEP = STEP / SORT10
C
DIAGONAL ELEMENTS. THREE POINT DIFFERENCING.
DENOM = STEP * STEP
DO 30 I = 1, NVAR
IND = ISUB(I)
PAR1 = PAR(IND)
PAR(IND) = PAR1 + STEP
CALL FUNCT(0, NPAR, PAR, TEMP, PAR, IW, LIW, W, LW)
PAR(IND) = PAR1 - STEP
CALL FUNCT(0, NPAR, PAR, TEMP1, PAR, IW, LIW, W, LW)
TEMP = TEMP + TEMP1
PAR(IND) = PAR1
CALL FUNCT(0, NPAR, PAR, TEMP1, PAR, IW, LIW, W, LW)
H(I,I) = (TEMP - TWO*TEMP1) / DENOM
30 CONTINUE
IF (NVAR .EQ. 1) GO TO 50
C
OFF-DIAGONAL ELEMENTS IF REQUIRED. FOUR POINT DIFFERENCING.
DENOM = FOUR * DENOM
IND = NVAR - 1
DO 40 I = 1, IND
IND1 = I + 1
IND2 = ISUB(I)
PAR1 = PAR(IND2)
DO 40 J = IND1, NVAR
IND3 = ISUB(J)
PARJ = PAR(IND3)
PAR(IND2) = PAR1 + STEP
PAR(IND3) = PARJ + STEP
CALL FUNCT(0, NPAR, PAR, TEMP, PAR, IW, LIW, W, LW)
PAR(IND3) = PARJ - STEP
CALL FUNCT(0, NPAR, PAR, TEMP1, PAR, IW, LIW, W, LW)
TEMP = TEMP - TEMP1
PAR(IND2) = PAR1 - STEP

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C      CALL FUNCT(O, NPAR, PAR, TEMP1, PAR, IW, LIW, W, LW)
      TEMP = TEMP + TEMP1
      PAR(IND3) = PARJ + STEP
      CALL FUNCT(O, NPAR, PAR, TEMP1, PAR, IW, LIW, W, LW)
      H(I,J) = (TEMP - TEMP1) / DENOM
      PAR(IND2) = PAR1
      PAR(IND3) = PARJ
40 CONTINUE
C
C      FIND DIFFERENCE, SAVE OLD HESSIAN IN SAVE1.
50 TEMP = ZERO
   DO 60 I = 1, NVAR
   DO 60 J = 1, NVAR
      PAR1 = H(I,J)
      TEMP1 = ABS(PAR1 - SAVE1(I,J))
      PARJ = ABS(PAR1)
      IF (PARJ .GE. ONE) TEMP1 = TEMP1 / PARJ
      TEMP = AMAX1(TEMP, TEMP1)
      SAVE1(I,J) = PAR1
60 CONTINUE
C
C      THIRD ITM, SAVE OLD HESSIAN IN SAVE2 (STEPLNGTH 1.0E-4).
   IF (ITER .NE. 3) GO TO 80
   DO 70 I = 1, NVAR
   DO 70 J = 1, NVAR
      DO 70 SAVE2(I,J) = H(I,J)
70 CONTINUE
C
C      TEST STOPPING CRITERION FOR ITERATION.
80 IF (TEMP .LT. TOL) GO TO 100
   IF (ITER .LT. 5) GO TO 20
C
C      NO CONVERGENCE IN 5 ITNS - USE SAVE2, WITH STEP TENM4.
      STEP = TENM4
   DO 90 I = 1, NVAR
   DO 90 J = 1, NVAR
      90 H(I,J) = SAVE2(I,J)
C
C      EXIT, WRITE DETAILS (IND IS OUTPUT UNIT PASSED THROUGH IW),
C      AND SAVE THE NUMBER OF ITERATIONS REQUIRED.
100 IND = IW(4)
   IW(1) = ITER
   IF (IND .GT. 0) WRITE (IND,1000) ITER, STEP
C
C      1000 FORMAT (16HONDISSIAN DONE IN, I3, 6H ITNS,, 16H STEPSIZE USED =,
C      *E11.3)
C
C      RETURN
      END

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20. ABSTRACT (Continue on reverse side if necessary and identify by block number) This paper presents several families of algorithms for estimation of the parameters of the stable laws and the parameters of attracting stable laws. The paper also presents algorithms for estimation of the parameters of stable regression and stable autoregression models.		

